

# Project

```
library(car)
```

```
## Warning: package 'car' was built under R version 4.0.5
```

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

```
library(randomForest)
```

```
## Warning: package 'randomForest' was built under R version 4.0.4
```

```
## randomForest 4.6-14
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
library(glmnet)
```

```
## Warning: package 'glmnet' was built under R version 4.0.5
```

```
## Loading required package: Matrix
```

```
## Loaded glmnet 4.1-3
```

```
library(pls)
```

```
## Warning: package 'pls' was built under R version 4.0.5
```

```
##  
## Attaching package: 'pls'
```

```
## The following object is masked from 'package:stats':  
##  
##   loadings
```

```
library(mgcv)
```

```
## Loading required package: nlme
```

```
## This is mgcv 1.8-33. For overview type 'help("mgcv-package")'.
```

```
data <- read.csv("kc_house_data.csv", header = TRUE)
head(data)
```

```
##           id           date    price bedrooms bathrooms sqft_living sqft_lot
## 1 7129300520 20141013T000000 221900         3         1.00       1180     5650
## 2 6414100192 20141209T000000 538000         3         2.25       2570     7242
## 3 5631500400 20150225T000000 180000         2         1.00        770    10000
## 4 2487200875 20141209T000000 604000         4         3.00       1960     5000
## 5 1954400510 20150218T000000 510000         3         2.00       1680     8080
## 6 7237550310 20140512T000000 1225000        4         4.50       5420    101930
##  floors waterfront view condition grade sqft_above sqft_basement yr_built
## 1         1           0      0         3         7       1180           0    1955
## 2         2           0      0         3         7       2170          400    1951
## 3         1           0      0         3         6        770           0    1933
## 4         1           0      0         5         7       1050          910    1965
## 5         1           0      0         3         8       1680           0    1987
## 6         1           0      0         3        11       3890         1530    2001
##  yr_renovated zipcode      lat      long sqft_living15 sqft_lot15
## 1           0    98178 47.5112 -122.257       1340       5650
## 2          1991    98125 47.7210 -122.319       1690       7639
## 3           0    98028 47.7379 -122.233       2720       8062
## 4           0    98136 47.5208 -122.393       1360       5000
## 5           0    98074 47.6168 -122.045       1800       7503
## 6           0    98053 47.6561 -122.005       4760      101930
```

```
dim(data)
```

```
## [1] 21613    21
```

```
# Summary statistics of the data
summary(data)
```

```
##          id          date          price          bedrooms
## Min.      :1.000e+06  Length:21613    Min.       : 75000  Min.       : 0.000
## 1st Qu.:2.123e+09    Class :character 1st Qu.: 321950    1st Qu.: 3.000
## Median :3.905e+09    Mode  :character Median : 450000    Median : 3.000
## Mean     :4.580e+09                                Mean     : 540088    Mean     : 3.371
## 3rd Qu.:7.309e+09                                3rd Qu.: 645000    3rd Qu.: 4.000
## Max.     :9.900e+09                                Max.     :770000    Max.     :33.000
##   bathrooms    sqft_living    sqft_lot    floors
## Min.       :0.000    Min.       : 290    Min.       : 520    Min.       :1.000
## 1st Qu.:1.750    1st Qu.: 1427    1st Qu.: 5040    1st Qu.:1.000
## Median :2.250    Median : 1910    Median : 7618    Median :1.500
## Mean     :2.115    Mean     : 2080    Mean     : 15107    Mean     :1.494
## 3rd Qu.:2.500    3rd Qu.: 2550    3rd Qu.: 10688    3rd Qu.:2.000
## Max.     :8.000    Max.     :13540    Max.     :1651359    Max.     :3.500
##   waterfront    view    condition    grade
## Min.       :0.000000    Min.       :0.0000    Min.       :1.000    Min.       : 1.000
## 1st Qu.:0.000000    1st Qu.:0.0000    1st Qu.:3.000    1st Qu.: 7.000
## Median :0.000000    Median :0.0000    Median :3.000    Median : 7.000
## Mean     :0.007542    Mean     :0.2343    Mean     :3.409    Mean     : 7.657
## 3rd Qu.:0.000000    3rd Qu.:0.0000    3rd Qu.:4.000    3rd Qu.: 8.000
## Max.     :1.000000    Max.     :4.0000    Max.     :5.000    Max.     :13.000
##   sqft_above    sqft_basement    yr_built    yr_renovated
## Min.       : 290    Min.       : 0.0    Min.       :1900    Min.       : 0.0
## 1st Qu.:1190    1st Qu.: 0.0    1st Qu.:1951    1st Qu.: 0.0
## Median :1560    Median : 0.0    Median :1975    Median : 0.0
## Mean     :1788    Mean     : 291.5    Mean     :1971    Mean     : 84.4
## 3rd Qu.:2210    3rd Qu.: 560.0    3rd Qu.:1997    3rd Qu.: 0.0
## Max.     :9410    Max.     :4820.0    Max.     :2015    Max.     :2015.0
##   zipcode    lat    long    sqft_living15
## Min.       :98001    Min.       :47.16    Min.       : -122.5    Min.       : 399
## 1st Qu.:98033    1st Qu.:47.47    1st Qu.: -122.3    1st Qu.:1490
## Median :98065    Median :47.57    Median : -122.2    Median :1840
## Mean     :98078    Mean     :47.56    Mean     : -122.2    Mean     :1987
## 3rd Qu.:98118    3rd Qu.:47.68    3rd Qu.: -122.1    3rd Qu.:2360
## Max.     :98199    Max.     :47.78    Max.     : -121.3    Max.     :6210
##   sqft_lot15
## Min.       : 651
## 1st Qu.: 5100
## Median : 7620
## Mean     : 12768
## 3rd Qu.: 10083
## Max.     :871200
```

```
# Determining whether any data is missing
sum(is.na(data))
```

```
## [1] 0
```

```
# Investigating the 33 bedroom property as this seems to be an outlier
data[data$bedrooms==33,]
```

```
##           id           date  price bedrooms bathrooms sqft_living sqft_lot
## 15871 2402100895 20140625T000000 640000         33         1.75         1620         6000
##      floors waterfront view condition grade sqft_above sqft_basement yr_built
## 15871      1          0    0          5      7         1040          580         1947
##      yr_renovated zipcode    lat    long sqft_living15 sqft_lot15
## 15871           0    98103 47.6878 -122.331         1330          4700
```

```
data[data$bedrooms == 0,]
```

##	id	date	price	bedrooms	bathrooms	sqft_living		
## 876	6306400140	20140612T000000	1095000	0	0.00	3064		
## 3120	3918400017	20150205T000000	380000	0	0.00	1470		
## 3468	1453602309	20140805T000000	288000	0	1.50	1430		
## 4869	6896300380	20141002T000000	228000	0	1.00	390		
## 6995	2954400190	20140624T000000	1295650	0	0.00	4810		
## 8478	2569500210	20141117T000000	339950	0	2.50	2290		
## 8485	2310060040	20140925T000000	240000	0	2.50	1810		
## 9774	3374500520	20150429T000000	355000	0	0.00	2460		
## 9855	7849202190	20141223T000000	235000	0	0.00	1470		
## 12654	7849202299	20150218T000000	320000	0	2.50	1490		
## 14424	9543000205	20150413T000000	139950	0	0.00	844		
## 18380	1222029077	20141029T000000	265000	0	0.75	384		
## 19453	3980300371	20140926T000000	142000	0	0.00	290		
##	sqft_lot	floors	waterfront	view	condition	grade	sqft_above	sqft_basement
## 876	4764	3.5	0	2	3	7	3064	0
## 3120	979	3.0	0	2	3	8	1470	0
## 3468	1650	3.0	0	0	3	7	1430	0
## 4869	5900	1.0	0	0	2	4	390	0
## 6995	28008	2.0	0	0	3	12	4810	0
## 8478	8319	2.0	0	0	3	8	2290	0
## 8485	5669	2.0	0	0	3	7	1810	0
## 9774	8049	2.0	0	0	3	8	2460	0
## 9855	4800	2.0	0	0	3	7	1470	0
## 12654	7111	2.0	0	0	3	7	1490	0
## 14424	4269	1.0	0	0	4	7	844	0
## 18380	213444	1.0	0	0	3	4	384	0
## 19453	20875	1.0	0	0	1	1	290	0
##	yr_built	yr_renovated	zipcode	lat	long	sqft_living15	sqft_lot15	
## 876	1990	0	98102	47.6362	-122.322	2360	4000	
## 3120	2006	0	98133	47.7145	-122.356	1470	1399	
## 3468	1999	0	98125	47.7222	-122.290	1430	1650	
## 4869	1953	0	98118	47.5260	-122.261	2170	6000	
## 6995	1990	0	98053	47.6642	-122.069	4740	35061	
## 8478	1985	0	98042	47.3473	-122.151	2500	8751	
## 8485	2003	0	98038	47.3493	-122.053	1810	5685	
## 9774	1990	0	98031	47.4095	-122.168	2520	8050	
## 9855	1996	0	98065	47.5265	-121.828	1060	7200	
## 12654	1999	0	98065	47.5261	-121.826	1500	4675	
## 14424	1913	0	98001	47.2781	-122.250	1380	9600	
## 18380	2003	0	98070	47.4177	-122.491	1920	224341	
## 19453	1963	0	98024	47.5308	-121.888	1620	22850	

```
data[data$bathrooms == 0, ]
```

##	id	date	price	bedrooms	bathrooms	sqft_living		
## 876	6306400140	20140612T000000	1095000	0	0	3064		
## 1150	3421079032	20150217T000000	75000	1	0	670		
## 3120	3918400017	20150205T000000	380000	0	0	1470		
## 5833	5702500050	20141104T000000	280000	1	0	600		
## 6995	2954400190	20140624T000000	1295650	0	0	4810		
## 9774	3374500520	20150429T000000	355000	0	0	2460		
## 9855	7849202190	20141223T000000	235000	0	0	1470		
## 10482	203100435	20140918T000000	484000	1	0	690		
## 14424	9543000205	20150413T000000	139950	0	0	844		
## 19453	3980300371	20140926T000000	142000	0	0	290		
##	sqft_lot	floors	waterfront	view	condition	grade	sqft_above	sqft_basement
## 876	4764	3.5	0	2	3	7	3064	0
## 1150	43377	1.0	0	0	3	3	670	0
## 3120	979	3.0	0	2	3	8	1470	0
## 5833	24501	1.0	0	0	2	3	600	0
## 6995	28008	2.0	0	0	3	12	4810	0
## 9774	8049	2.0	0	0	3	8	2460	0
## 9855	4800	2.0	0	0	3	7	1470	0
## 10482	23244	1.0	0	0	4	7	690	0
## 14424	4269	1.0	0	0	4	7	844	0
## 19453	20875	1.0	0	0	1	1	290	0
##	yr_built	yr_renovated	zipcode	lat	long	sqft_living15	sqft_lot15	
## 876	1990	0	98102	47.6362	-122.322	2360	4000	
## 1150	1966	0	98022	47.2638	-121.906	1160	42882	
## 3120	2006	0	98133	47.7145	-122.356	1470	1399	
## 5833	1950	0	98045	47.5316	-121.749	990	22549	
## 6995	1990	0	98053	47.6642	-122.069	4740	35061	
## 9774	1990	0	98031	47.4095	-122.168	2520	8050	
## 9855	1996	0	98065	47.5265	-121.828	1060	7200	
## 10482	1948	0	98053	47.6429	-121.955	1690	19290	
## 14424	1913	0	98001	47.2781	-122.250	1380	9600	
## 19453	1963	0	98024	47.5308	-121.888	1620	22850	

```
# Removing properties that are unusual (33 bedrooms, no bedrooms or bathrooms)
data <- data[-which.max(data$bedrooms),]
data <- data[-which(data$bedrooms == 0),]
data <- data[-which(data$bathrooms == 0),]

# Removing the ID column as this is just a unique identifier for each property
data <- data[,-1]

# Converting categorical variables to factors
data$zipcode <- as.factor(data$zipcode)

# Converting to price per square foot
data$price_sqft <- data$price / data$sqft_living
data <- data[,c(-2, -5)]

data$waterfront <- as.factor(data$waterfront)

summary(data)
```

```

##      date      bedrooms      bathrooms      sqft_lot
## Length:21596      Min.   : 1.000      Min.   :0.500      Min.   : 520
## Class :character  1st Qu.: 3.000      1st Qu.:1.750      1st Qu.: 5040
## Mode  :character  Median : 3.000      Median :2.250      Median : 7619
##                               Mean  : 3.372      Mean   :2.116      Mean   : 15100
##                               3rd Qu.: 4.000      3rd Qu.:2.500      3rd Qu.: 10686
##                               Max.   :11.000      Max.   :8.000      Max.   :1651359
##
##      floors      waterfront      view      condition      grade
## Min.   :1.000      0:21433      Min.   :0.0000      Min.   :1.00      Min.   : 3.000
## 1st Qu.:1.000      1: 163      1st Qu.:0.0000      1st Qu.:3.00      1st Qu.: 7.000
## Median :1.500                               Median :0.0000      Median :3.00      Median : 7.000
## Mean   :1.494                               Mean   :0.2343      Mean   :3.41      Mean   : 7.658
## 3rd Qu.:2.000                               3rd Qu.:0.0000      3rd Qu.:4.00      3rd Qu.: 8.000
## Max.   :3.500                               Max.   :4.0000      Max.   :5.00      Max.   :13.000
##
##      sqft_above      sqft_basement      yr_built      yr_renovated
## Min.   : 370      Min.   : 0.0      Min.   :1900      Min.   : 0.00
## 1st Qu.:1190      1st Qu.: 0.0      1st Qu.:1951      1st Qu.: 0.00
## Median :1560      Median : 0.0      Median :1975      Median : 0.00
## Mean   :1789      Mean   :291.7      Mean   :1971      Mean   : 84.47
## 3rd Qu.:2210      3rd Qu.:560.0      3rd Qu.:1997      3rd Qu.: 0.00
## Max.   :9410      Max.   :4820.0      Max.   :2015      Max.   :2015.00
##
##      zipcode      lat      long      sqft_living15
## 98103 : 601      Min.   :47.16      Min.   :-122.5      Min.   : 399
## 98038 : 589      1st Qu.:47.47      1st Qu.: -122.3      1st Qu.:1490
## 98115 : 583      Median :47.57      Median : -122.2      Median :1840
## 98052 : 574      Mean   :47.56      Mean   : -122.2      Mean   :1987
## 98117 : 553      3rd Qu.:47.68      3rd Qu.: -122.1      3rd Qu.:2360
## 98042 : 547      Max.   :47.78      Max.   : -121.3      Max.   :6210
## (Other):18149
##      sqft_lot15      price_sqft
## Min.   : 651      Min.   : 87.59
## 1st Qu.: 5100      1st Qu.:182.29
## Median : 7620      Median :244.63
## Mean   : 12759      Mean   :264.11
## 3rd Qu.: 10083      3rd Qu.:318.27
## Max.   :871200      Max.   :810.14
##

```



```

# Converting the date data to dates in R
Dates <- NULL
for (i in data$date) {
  # Grabbing the date in format yyyymmdd
  d <- substr(i, 1, 8)
  Dates <- rbind(Dates, d)
}

# Converting strings to dates
Dates <- as.Date(Dates, "%Y%m%d")

# Updating the date column to the new date values
data$date <- Dates

head(data$date)

```

```

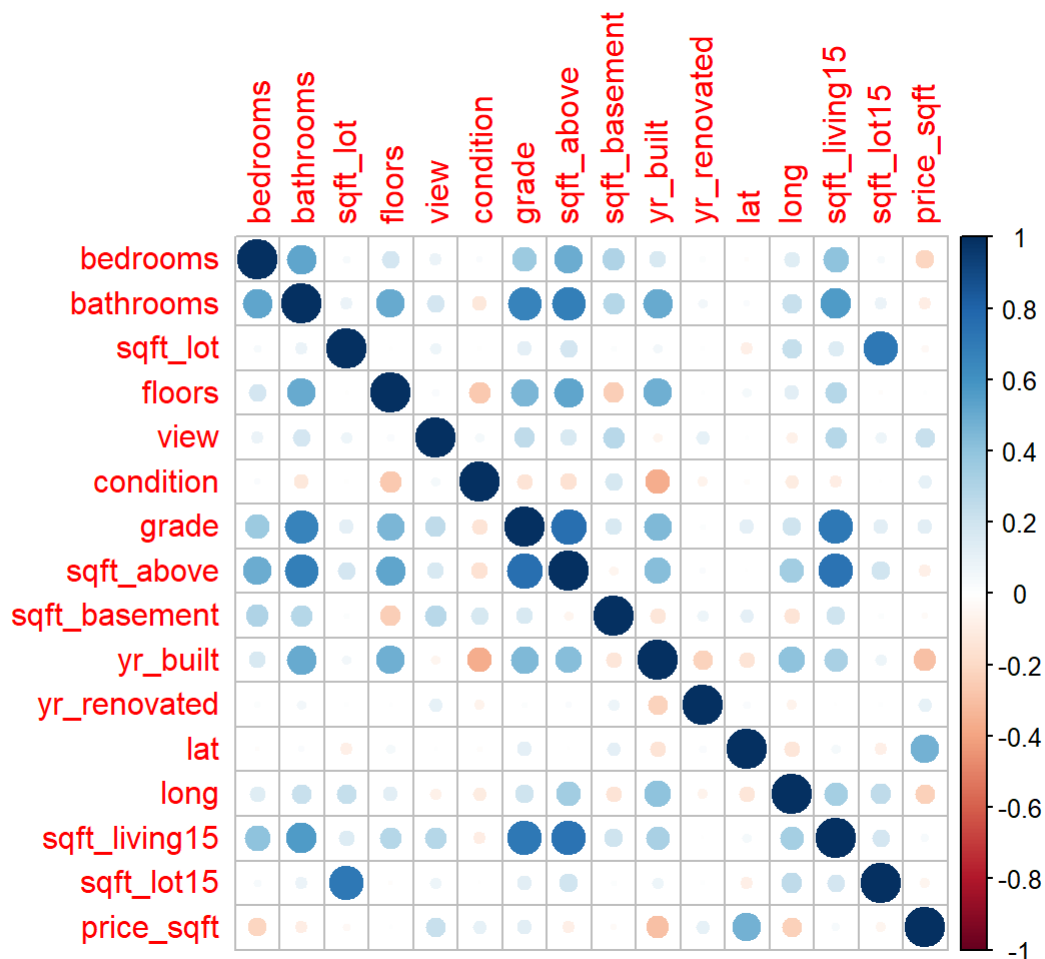
## [1] "2014-10-13" "2014-12-09" "2015-02-25" "2014-12-09" "2015-02-18"
## [6] "2014-05-12"

```

```

# Correlation between the numeric predictors
corrplot::corrplot(cor(data[,c(-1, -6, -14)]))

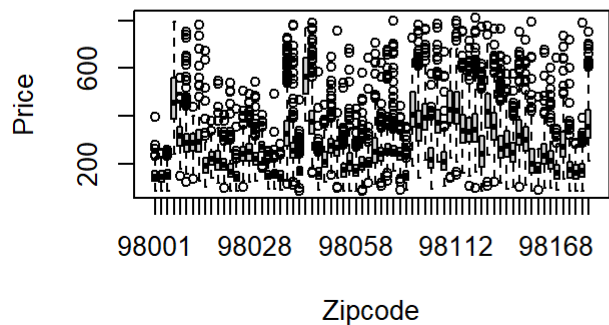
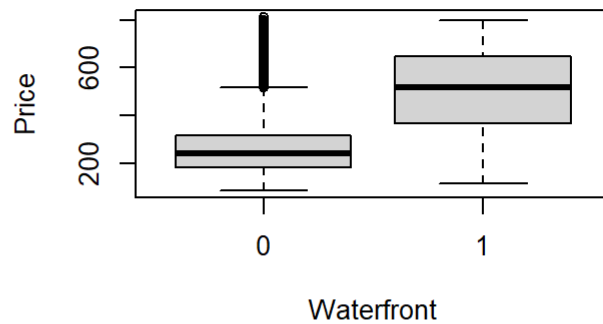
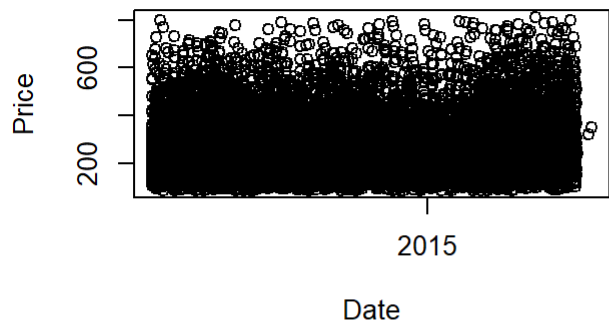
```



```

par(mfrow=c(2,2))
plot(data$date, data$price, xlab = "Date", ylab = "Price")
plot(data$waterfront, data$price, xlab = "Waterfront", ylab = "Price")
plot(data$zipcode, data$price, xlab = "Zipcode", ylab = "Price")

```

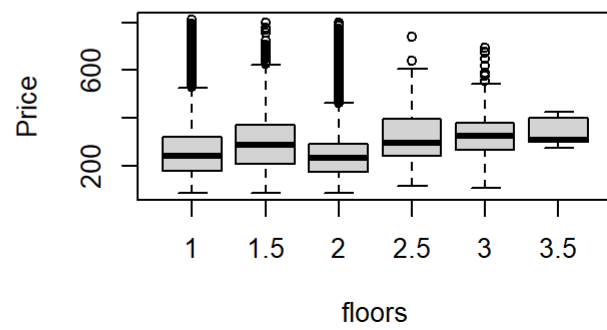
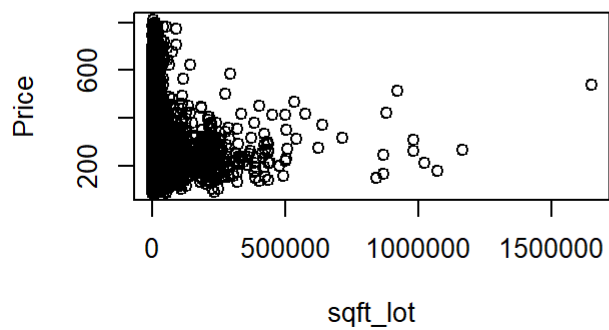
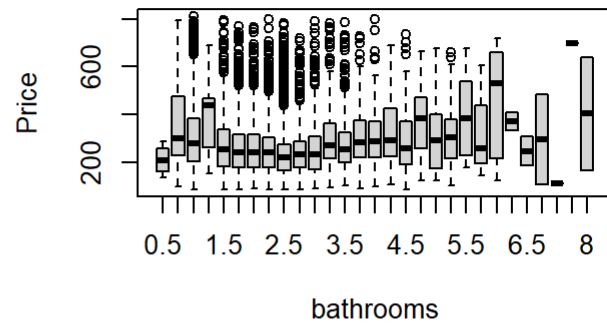
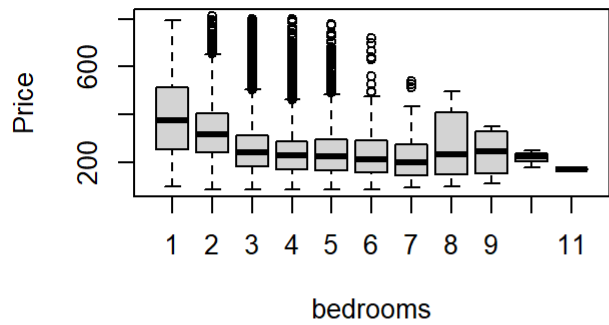


```

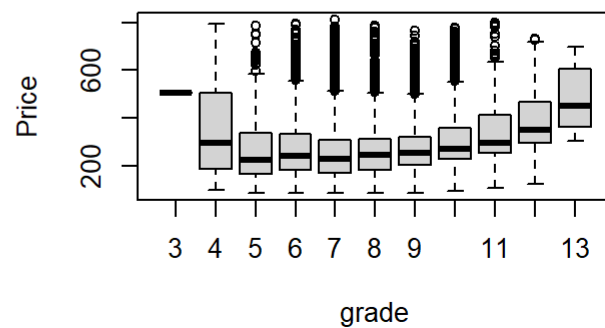
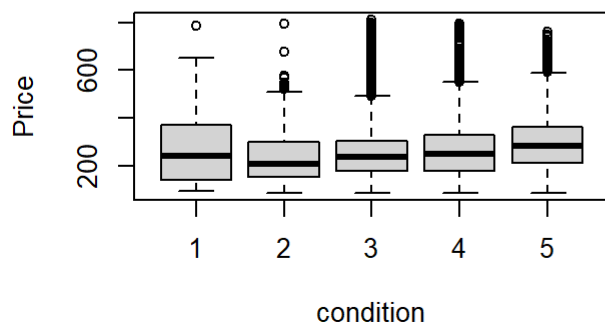
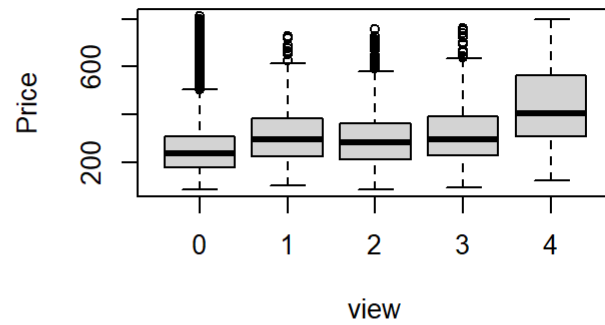
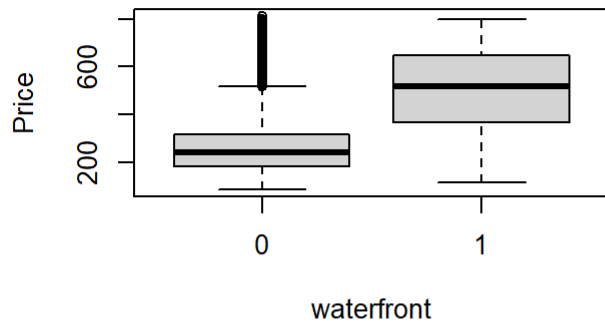
num.cols <- colnames(data[,c(-1, -14)])

par(mfrow=c(2,2))
for (i in num.cols[1:4]) {
  if (i == num.cols[3]) {
    plot(data[,i], data$price, xlab = i, ylab = "Price")
  } else {
    plot(as.factor(data[,i]), data$price, xlab = i, ylab = "Price")
  }
}

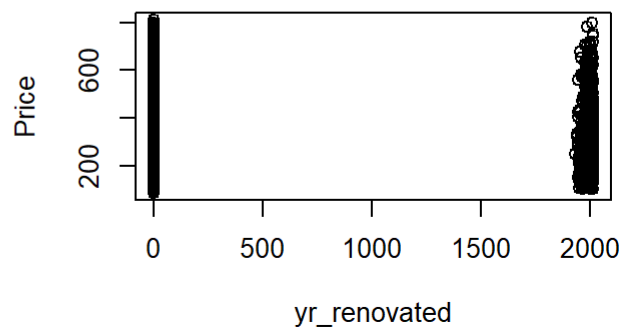
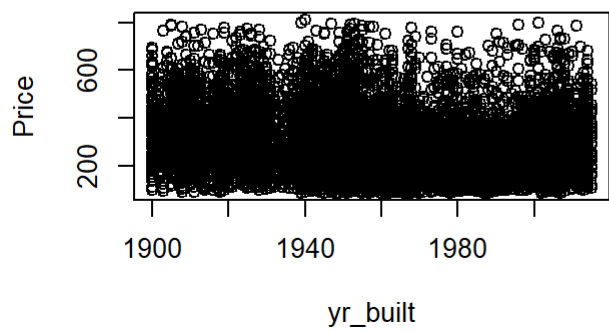
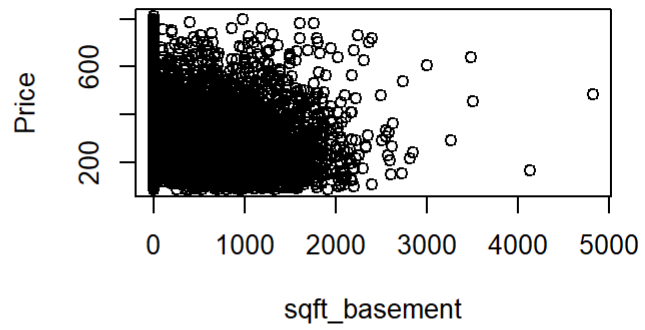
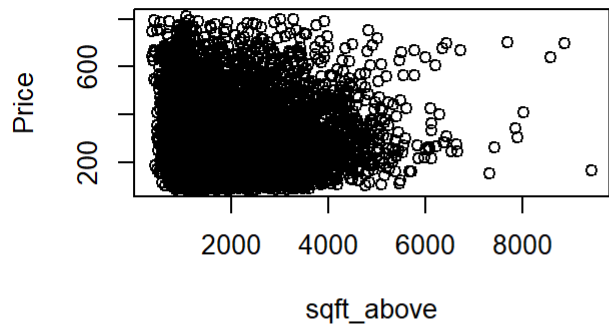
```



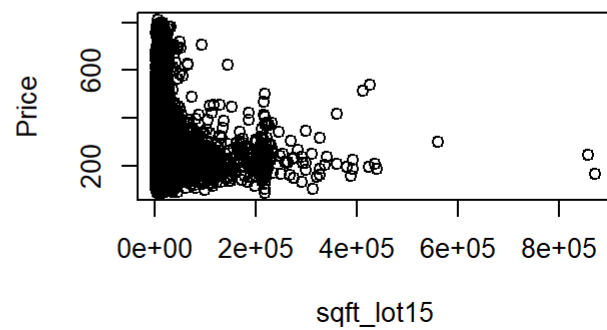
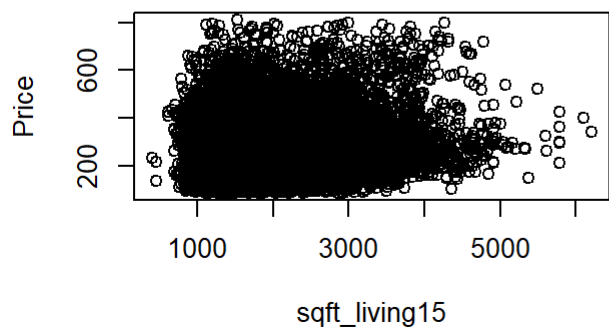
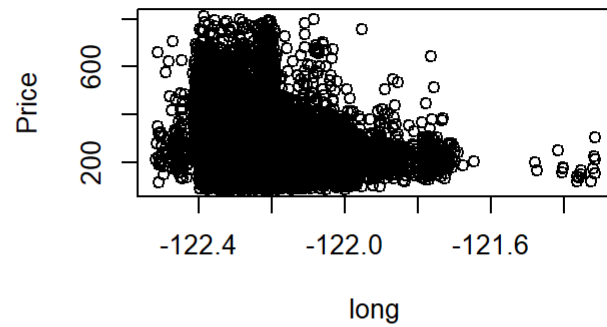
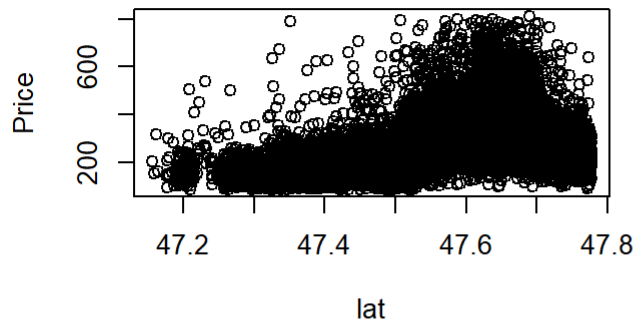
```
par(mfrow=c(2,2))
for (i in num.cols[5:8]) {
  plot(as.factor(data[,i]), data$price, xlab = i, ylab = "Price")
}
```



```
par(mfrow=c(2,2))
for (i in num.cols[9:12]) {
  plot(data[,i], data$price, xlab = i, ylab = "Price")
}
```



```
par(mfrow=c(2,2))  
for (i in num.cols[13:16]) {  
  plot(data[,i], data$price, xlab = i, ylab = "Price")  
}
```



```
# Building the full model
```

```
lm1 <- lm(price_sqft ~ ., data = data)  
summary(lm1)
```

```
##
## Call:
## lm(formula = price_sqft ~ ., data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -344.37  -30.92   -3.56   23.75  564.97
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.555e+04  2.245e+03  -6.924 4.51e-12 ***
## date         6.210e-02  3.555e-03  17.467 < 2e-16 ***
## bedrooms    -1.318e+01  5.867e-01 -22.463 < 2e-16 ***
## bathrooms    4.360e+00  9.644e-01   4.521 6.18e-06 ***
## sqft_lot     1.738e-04  1.404e-05  12.380 < 2e-16 ***
## floors      -1.946e+01  1.156e+00 -16.830 < 2e-16 ***
## waterfront1  1.972e+02  5.166e+00  38.173 < 2e-16 ***
## view         1.964e+01  6.416e-01  30.613 < 2e-16 ***
## condition    1.019e+01  7.057e-01  14.432 < 2e-16 ***
## grade        1.853e+01  6.668e-01  27.785 < 2e-16 ***
## sqft_above   -3.769e-02  1.111e-03 -33.918 < 2e-16 ***
## sqft_basement -7.571e-02  1.303e-03 -58.118 < 2e-16 ***
## yr_built     -2.491e-01  2.376e-02 -10.481 < 2e-16 ***
## yr_renovated  6.515e-03  1.077e-03   6.050 1.47e-09 ***
## zipcode98002  3.456e+00  5.287e+00   0.654 0.513295
## zipcode98003 -5.960e+00  4.727e+00  -1.261 0.207366
## zipcode98004  3.072e+02  8.590e+00  35.756 < 2e-16 ***
## zipcode98005  1.466e+02  9.183e+00  15.970 < 2e-16 ***
## zipcode98006  1.335e+02  7.508e+00  17.783 < 2e-16 ***
## zipcode98007  1.255e+02  9.476e+00  13.245 < 2e-16 ***
## zipcode98008  1.278e+02  9.001e+00  14.197 < 2e-16 ***
## zipcode98010  7.238e+01  8.060e+00   8.979 < 2e-16 ***
## zipcode98011  5.167e+01  1.171e+01   4.413 1.02e-05 ***
## zipcode98014  6.823e+01  1.286e+01   5.304 1.14e-07 ***
## zipcode98019  5.327e+01  1.269e+01   4.199 2.70e-05 ***
## zipcode98022  2.850e+01  7.004e+00   4.069 4.74e-05 ***
## zipcode98023 -1.619e+01  4.349e+00  -3.722 0.000198 ***
## zipcode98024  9.702e+01  1.132e+01   8.572 < 2e-16 ***
## zipcode98027  1.097e+02  7.706e+00  14.240 < 2e-16 ***
## zipcode98028  4.333e+01  1.137e+01   3.810 0.000140 ***
## zipcode98029  1.262e+02  8.804e+00  14.337 < 2e-16 ***
## zipcode98030  8.807e+00  5.198e+00   1.694 0.090231 .
## zipcode98031  1.133e+01  5.414e+00   2.092 0.036459 *
## zipcode98032 -6.628e+00  6.283e+00  -1.055 0.291485
## zipcode98033  1.664e+02  9.756e+00  17.053 < 2e-16 ***
## zipcode98034  8.474e+01  1.046e+01   8.101 5.75e-16 ***
## zipcode98038  4.329e+01  5.837e+00   7.416 1.25e-13 ***
## zipcode98039  4.004e+02  1.161e+01  34.498 < 2e-16 ***
## zipcode98040  2.076e+02  7.598e+00  27.329 < 2e-16 ***
## zipcode98042  1.973e+01  4.975e+00   3.966 7.33e-05 ***
## zipcode98045  9.305e+01  1.078e+01   8.628 < 2e-16 ***
## zipcode98052  1.145e+02  9.960e+00  11.497 < 2e-16 ***
```

```

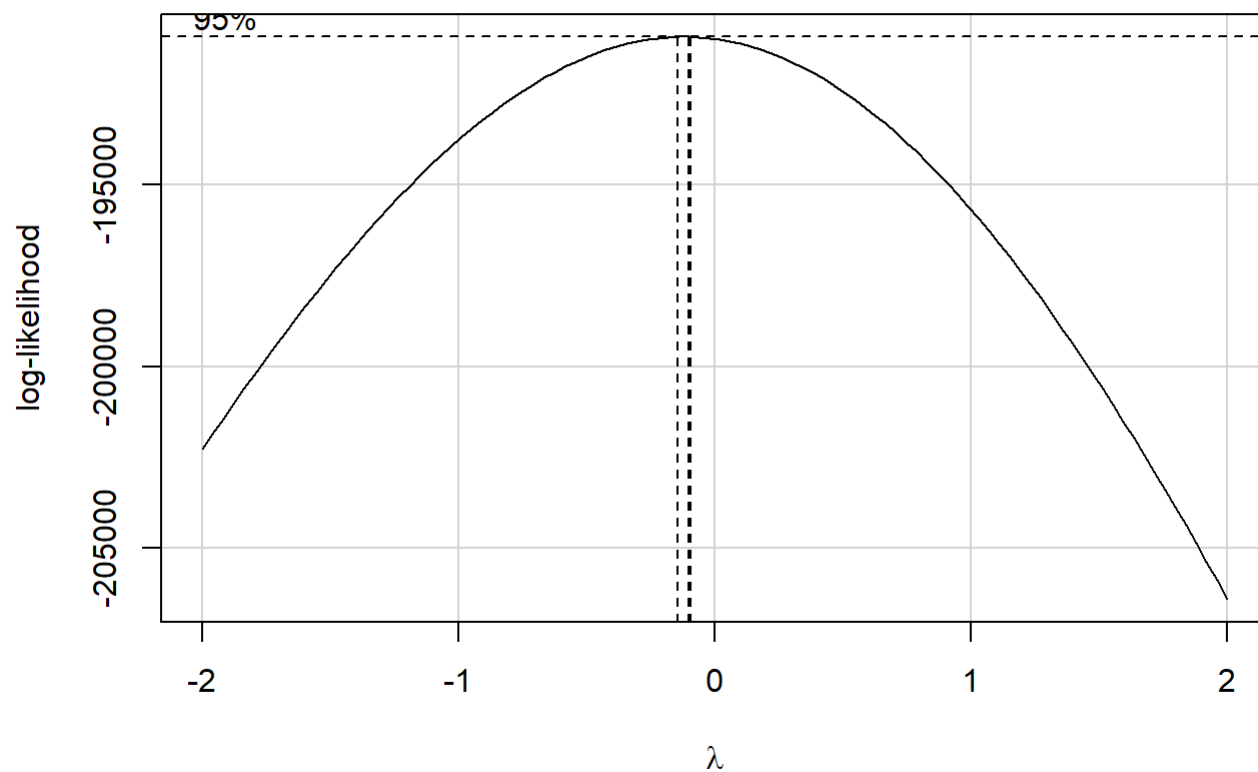
## zipcode98053 1.080e+02 1.067e+01 10.116 < 2e-16 ***
## zipcode98055 1.854e+01 6.029e+00 3.075 0.002105 **
## zipcode98056 5.373e+01 6.550e+00 8.203 2.47e-16 ***
## zipcode98058 2.649e+01 5.696e+00 4.650 3.33e-06 ***
## zipcode98059 5.860e+01 6.426e+00 9.119 < 2e-16 ***
## zipcode98065 8.977e+01 9.942e+00 9.029 < 2e-16 ***
## zipcode98070 1.336e+01 7.591e+00 1.760 0.078442 .
## zipcode98072 6.987e+01 1.165e+01 5.999 2.02e-09 ***
## zipcode98074 1.055e+02 9.432e+00 11.187 < 2e-16 ***
## zipcode98075 1.097e+02 9.070e+00 12.099 < 2e-16 ***
## zipcode98077 6.872e+01 1.212e+01 5.670 1.45e-08 ***
## zipcode98092 9.197e+00 4.727e+00 1.946 0.051723 .
## zipcode98102 2.401e+02 1.007e+01 23.850 < 2e-16 ***
## zipcode98103 1.763e+02 9.434e+00 18.684 < 2e-16 ***
## zipcode98105 2.159e+02 9.685e+00 22.289 < 2e-16 ***
## zipcode98106 5.525e+01 6.988e+00 7.907 2.77e-15 ***
## zipcode98107 1.846e+02 9.725e+00 18.977 < 2e-16 ***
## zipcode98108 5.568e+01 7.716e+00 7.215 5.56e-13 ***
## zipcode98109 2.353e+02 1.002e+01 23.484 < 2e-16 ***
## zipcode98112 2.603e+02 8.891e+00 29.276 < 2e-16 ***
## zipcode98115 1.663e+02 9.589e+00 17.343 < 2e-16 ***
## zipcode98116 1.494e+02 7.803e+00 19.143 < 2e-16 ***
## zipcode98117 1.647e+02 9.711e+00 16.956 < 2e-16 ***
## zipcode98118 8.552e+01 6.816e+00 12.548 < 2e-16 ***
## zipcode98119 2.331e+02 9.464e+00 24.630 < 2e-16 ***
## zipcode98122 1.793e+02 8.443e+00 21.231 < 2e-16 ***
## zipcode98125 8.862e+01 1.036e+01 8.556 < 2e-16 ***
## zipcode98126 9.821e+01 7.166e+00 13.706 < 2e-16 ***
## zipcode98133 5.668e+01 1.069e+01 5.301 1.17e-07 ***
## zipcode98136 1.341e+02 7.346e+00 18.262 < 2e-16 ***
## zipcode98144 1.326e+02 7.848e+00 16.897 < 2e-16 ***
## zipcode98146 3.411e+01 6.558e+00 5.201 2.00e-07 ***
## zipcode98148 1.183e+01 8.920e+00 1.327 0.184654
## zipcode98155 4.839e+01 1.112e+01 4.352 1.36e-05 ***
## zipcode98166 3.135e+01 6.002e+00 5.223 1.77e-07 ***
## zipcode98168 3.029e+00 6.343e+00 0.477 0.633012
## zipcode98177 8.181e+01 1.116e+01 7.329 2.40e-13 ***
## zipcode98178 1.459e+01 6.553e+00 2.227 0.025945 *
## zipcode98188 8.551e+00 6.725e+00 1.272 0.203535
## zipcode98198 -6.999e+00 5.096e+00 -1.374 0.169575
## zipcode98199 1.760e+02 9.219e+00 19.095 < 2e-16 ***
## lat 7.763e+01 2.318e+01 3.349 0.000812 ***
## long -9.349e+01 1.665e+01 -5.617 1.97e-08 ***
## sqft_living15 2.098e-02 1.056e-03 19.858 < 2e-16 ***
## sqft_lot15 -1.130e-06 2.211e-05 -0.051 0.959250
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 58.71 on 21509 degrees of freedom
## Multiple R-squared:  0.716, Adjusted R-squared:  0.7149
## F-statistic: 630.7 on 86 and 21509 DF, p-value: < 2.2e-16

```



```
bc <- boxCox(lm1)
```

### Profile Log-likelihood



```
opt_lambda <- bc$x[which.max(bc$y)]
```

```
# Log transformation of dependent variable  
  
lm2 <- lm(log(price_sqft) ~ ., data = data)  
summary(lm2)
```

```
##
## Call:
## lm(formula = log(price_sqft) ~ ., data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.19773 -0.10794 -0.00113  0.10475  1.60278
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -5.465e+01  7.398e+00  -7.387 1.56e-13 ***
## date          2.097e-04  1.172e-05  17.901 < 2e-16 ***
## bedrooms     -4.504e-02  1.933e-03 -23.296 < 2e-16 ***
## bathrooms     2.007e-02  3.178e-03   6.314 2.76e-10 ***
## sqft_lot      6.820e-07  4.626e-08  14.742 < 2e-16 ***
## floors       -6.277e-02  3.811e-03 -16.470 < 2e-16 ***
## waterfront1   5.007e-01  1.703e-02  29.409 < 2e-16 ***
## view          6.652e-02  2.114e-03  31.462 < 2e-16 ***
## condition     4.704e-02  2.326e-03  20.228 < 2e-16 ***
## grade         7.928e-02  2.198e-03  36.076 < 2e-16 ***
## sqft_above    -1.555e-04  3.662e-06 -42.475 < 2e-16 ***
## sqft_basement -2.827e-04  4.293e-06 -65.862 < 2e-16 ***
## yr_built      -6.389e-04  7.832e-05  -8.158 3.59e-16 ***
## yr_renovated   2.873e-05  3.548e-06   8.097 5.92e-16 ***
## zipcode98002   8.574e-03  1.742e-02   0.492 0.622666
## zipcode98003  -2.222e-03  1.558e-02  -0.143 0.886573
## zipcode98004   1.019e+00  2.831e-02  35.990 < 2e-16 ***
## zipcode98005   6.121e-01  3.026e-02  20.227 < 2e-16 ***
## zipcode98006   5.762e-01  2.474e-02  23.287 < 2e-16 ***
## zipcode98007   5.413e-01  3.123e-02  17.335 < 2e-16 ***
## zipcode98008   5.440e-01  2.966e-02  18.338 < 2e-16 ***
## zipcode98010   3.159e-01  2.656e-02  11.894 < 2e-16 ***
## zipcode98011   2.390e-01  3.859e-02   6.194 5.96e-10 ***
## zipcode98014   2.596e-01  4.239e-02   6.124 9.30e-10 ***
## zipcode98019   2.007e-01  4.181e-02   4.802 1.59e-06 ***
## zipcode98022   1.538e-01  2.308e-02   6.664 2.74e-11 ***
## zipcode98023  -5.905e-02  1.433e-02  -4.120 3.80e-05 ***
## zipcode98024   4.181e-01  3.730e-02  11.208 < 2e-16 ***
## zipcode98027   4.854e-01  2.540e-02  19.116 < 2e-16 ***
## zipcode98028   2.036e-01  3.748e-02   5.432 5.64e-08 ***
## zipcode98029   5.513e-01  2.901e-02  19.003 < 2e-16 ***
## zipcode98030   3.301e-02  1.713e-02   1.927 0.054023 .
## zipcode98031   4.568e-02  1.784e-02   2.560 0.010472 *
## zipcode98032  -3.542e-02  2.071e-02  -1.711 0.087167 .
## zipcode98033   6.385e-01  3.215e-02  19.860 < 2e-16 ***
## zipcode98034   3.674e-01  3.447e-02  10.659 < 2e-16 ***
## zipcode98038   1.906e-01  1.924e-02   9.909 < 2e-16 ***
## zipcode98039   1.228e+00  3.825e-02  32.099 < 2e-16 ***
## zipcode98040   7.961e-01  2.504e-02  31.796 < 2e-16 ***
## zipcode98042   8.060e-02  1.640e-02   4.916 8.89e-07 ***
## zipcode98045   3.993e-01  3.554e-02  11.234 < 2e-16 ***
## zipcode98052   4.914e-01  3.282e-02  14.971 < 2e-16 ***
```

```

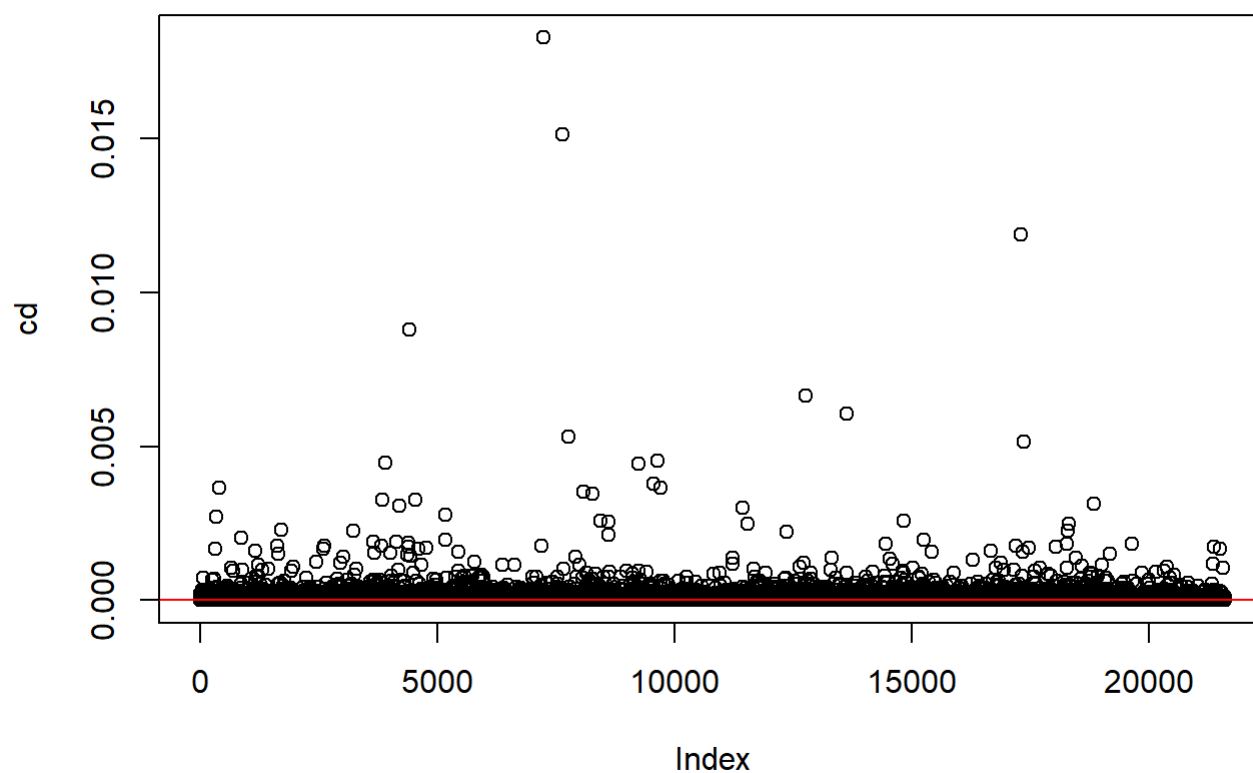
## zipcode98053 4.632e-01 3.517e-02 13.169 < 2e-16 ***
## zipcode98055 9.486e-02 1.987e-02 4.774 1.81e-06 ***
## zipcode98056 2.559e-01 2.159e-02 11.855 < 2e-16 ***
## zipcode98058 1.278e-01 1.877e-02 6.809 1.01e-11 ***
## zipcode98059 2.821e-01 2.118e-02 13.320 < 2e-16 ***
## zipcode98065 3.944e-01 3.276e-02 12.038 < 2e-16 ***
## zipcode98070 1.832e-01 2.502e-02 7.322 2.53e-13 ***
## zipcode98072 3.047e-01 3.838e-02 7.939 2.14e-15 ***
## zipcode98074 4.637e-01 3.108e-02 14.919 < 2e-16 ***
## zipcode98075 4.869e-01 2.989e-02 16.292 < 2e-16 ***
## zipcode98077 3.035e-01 3.994e-02 7.599 3.10e-14 ***
## zipcode98092 3.869e-02 1.558e-02 2.483 0.013017 *
## zipcode98102 8.434e-01 3.318e-02 25.419 < 2e-16 ***
## zipcode98103 6.572e-01 3.109e-02 21.139 < 2e-16 ***
## zipcode98105 7.833e-01 3.192e-02 24.543 < 2e-16 ***
## zipcode98106 2.723e-01 2.303e-02 11.822 < 2e-16 ***
## zipcode98107 6.841e-01 3.205e-02 21.346 < 2e-16 ***
## zipcode98108 2.602e-01 2.543e-02 10.231 < 2e-16 ***
## zipcode98109 8.189e-01 3.302e-02 24.802 < 2e-16 ***
## zipcode98112 8.998e-01 2.930e-02 30.709 < 2e-16 ***
## zipcode98115 6.416e-01 3.160e-02 20.304 < 2e-16 ***
## zipcode98116 6.098e-01 2.571e-02 23.716 < 2e-16 ***
## zipcode98117 6.267e-01 3.200e-02 19.584 < 2e-16 ***
## zipcode98118 3.779e-01 2.246e-02 16.827 < 2e-16 ***
## zipcode98119 8.029e-01 3.119e-02 25.746 < 2e-16 ***
## zipcode98122 6.787e-01 2.782e-02 24.394 < 2e-16 ***
## zipcode98125 3.808e-01 3.413e-02 11.156 < 2e-16 ***
## zipcode98126 4.503e-01 2.361e-02 19.068 < 2e-16 ***
## zipcode98133 2.609e-01 3.524e-02 7.403 1.38e-13 ***
## zipcode98136 5.657e-01 2.421e-02 23.372 < 2e-16 ***
## zipcode98144 5.573e-01 2.586e-02 21.549 < 2e-16 ***
## zipcode98146 1.891e-01 2.161e-02 8.749 < 2e-16 ***
## zipcode98148 9.729e-02 2.940e-02 3.310 0.000935 ***
## zipcode98155 2.292e-01 3.664e-02 6.254 4.08e-10 ***
## zipcode98166 2.147e-01 1.978e-02 10.855 < 2e-16 ***
## zipcode98168 2.663e-02 2.090e-02 1.274 0.202698
## zipcode98177 3.562e-01 3.678e-02 9.683 < 2e-16 ***
## zipcode98178 8.576e-02 2.159e-02 3.972 7.16e-05 ***
## zipcode98188 5.001e-02 2.216e-02 2.257 0.024041 *
## zipcode98198 1.873e-02 1.679e-02 1.115 0.264808
## zipcode98199 6.690e-01 3.038e-02 22.022 < 2e-16 ***
## lat 4.718e-01 7.638e-02 6.177 6.64e-10 ***
## long -2.847e-01 5.485e-02 -5.190 2.12e-07 ***
## sqft_living15 6.183e-05 3.481e-06 17.760 < 2e-16 ***
## sqft_lot15 1.069e-07 7.286e-08 1.467 0.142473
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1935 on 21509 degrees of freedom
## Multiple R-squared: 0.7605, Adjusted R-squared: 0.7596
## F-statistic: 794.2 on 86 and 21509 DF, p-value: < 2.2e-16

```

```
# Cook's distance to find outliers

cd <- cooks.distance(lm2)

plot(cd)
abline(h = quantile(cd, 4/nrow(data)), col = 'red')
```



```
# Removing outliers in the 99th percentile of cook's distance

#outliers <- which(cd > quantile(cd, .995))
outliers <- which(cd > 4/nrow(data))

data2 <- data[-outliers,]

summary(data2)
```

```
##      date      bedrooms      bathrooms      sqft_lot
## Min.   :2014-05-02   Min.    : 1.000   Min.    :0.500   Min.    : 520
## 1st Qu.:2014-07-21   1st Qu.: 3.000   1st Qu.:1.750   1st Qu.: 5001
## Median :2014-10-15   Median : 3.000   Median :2.250   Median : 7553
## Mean   :2014-10-28   Mean    : 3.378   Mean    :2.115   Mean    :13619
## 3rd Qu.:2015-02-17   3rd Qu.: 4.000   3rd Qu.:2.500   3rd Qu.:10440
## Max.   :2015-05-27   Max.    :10.000   Max.    :6.750   Max.    :982998
##
##      floors      waterfront      view      condition      grade
## Min.    :1.000    0:20277   Min.    :0.0000   Min.    :1.000   Min.    : 4.000
## 1st Qu.:1.000    1: 97     1st Qu.:0.0000   1st Qu.:3.000   1st Qu.: 7.000
## Median :1.500                Median :0.0000   Median :3.000   Median : 7.000
## Mean    :1.498                Mean    :0.2121   Mean    :3.415   Mean    : 7.661
## 3rd Qu.:2.000                3rd Qu.:0.0000   3rd Qu.:4.000   3rd Qu.: 8.000
## Max.    :3.500                Max.    :4.0000   Max.    :5.000   Max.    :13.000
##
##      sqft_above      sqft_basement      yr_built      yr_renovated
## Min.    : 390       Min.    : 0.0   Min.    :1900   Min.    : 0.00
## 1st Qu.:1200       1st Qu.: 0.0   1st Qu.:1953   1st Qu.: 0.00
## Median :1560       Median : 0.0   Median :1976   Median : 0.00
## Mean    :1779       Mean    :285.5   Mean    :1972   Mean    : 78.49
## 3rd Qu.:2200       3rd Qu.:550.0   3rd Qu.:1997   3rd Qu.: 0.00
## Max.    :7420       Max.    :3260.0   Max.    :2015   Max.    :2015.00
##
##      zipcode      lat      long      sqft_living15
## 98103 : 595       Min.    :47.16   Min.    :-122.5   Min.    : 460
## 98038 : 576       1st Qu.:47.47   1st Qu.: -122.3   1st Qu.:1490
## 98052 : 568       Median :47.57   Median : -122.2   Median :1840
## 98115 : 567       Mean    :47.56   Mean    : -122.2   Mean    :1983
## 98117 : 547       3rd Qu.:47.68   3rd Qu.: -122.1   3rd Qu.:2360
## 98042 : 536       Max.    :47.78   Max.    : -121.3   Max.    :5790
## (Other):16985
##      sqft_lot15      price_sqft
## Min.    : 651       Min.    : 88.08
## 1st Qu.: 5080       1st Qu.:183.49
## Median : 7565       Median :243.98
## Mean    :11926       Mean    :260.95
## 3rd Qu.: 9976       3rd Qu.:314.63
## Max.    :560617     Max.    :800.00
##
```

```
lm3 <- lm(log(price_sqft) ~ ., data = data2)
summary(lm3)
```

```
##
## Call:
## lm(formula = log(price_sqft) ~ ., data = data2)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-0.53336	-0.09760	-0.00049	0.09603	0.52377

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-3.902e+01	6.641e+00	-5.876	4.27e-09 ***
date	1.943e-04	9.542e-06	20.367	< 2e-16 ***
bedrooms	-3.274e-02	1.620e-03	-20.213	< 2e-16 ***
bathrooms	1.831e-02	2.643e-03	6.929	4.37e-12 ***
sqft_lot	8.735e-07	5.384e-08	16.225	< 2e-16 ***
floors	-6.210e-02	3.141e-03	-19.771	< 2e-16 ***
waterfront1	5.464e-01	1.681e-02	32.496	< 2e-16 ***
view	6.558e-02	1.776e-03	36.931	< 2e-16 ***
condition	4.421e-02	1.929e-03	22.919	< 2e-16 ***
grade	8.409e-02	1.834e-03	45.840	< 2e-16 ***
sqft_above	-1.851e-04	3.175e-06	-58.287	< 2e-16 ***
sqft_basement	-3.156e-04	3.658e-06	-86.265	< 2e-16 ***
yr_built	-5.645e-04	6.547e-05	-8.622	< 2e-16 ***
yr_renovated	3.448e-05	2.986e-06	11.549	< 2e-16 ***
zipcode98002	1.293e-02	1.419e-02	0.911	0.362388
zipcode98003	1.125e-02	1.262e-02	0.891	0.372848
zipcode98004	1.055e+00	2.341e-02	45.074	< 2e-16 ***
zipcode98005	6.570e-01	2.509e-02	26.182	< 2e-16 ***
zipcode98006	5.974e-01	2.066e-02	28.917	< 2e-16 ***
zipcode98007	5.784e-01	2.592e-02	22.319	< 2e-16 ***
zipcode98008	5.660e-01	2.481e-02	22.811	< 2e-16 ***
zipcode98010	2.520e-01	2.398e-02	10.511	< 2e-16 ***
zipcode98011	2.989e-01	3.179e-02	9.401	< 2e-16 ***
zipcode98014	2.717e-01	3.703e-02	7.337	2.26e-13 ***
zipcode98019	2.338e-01	3.549e-02	6.588	4.57e-11 ***
zipcode98022	1.174e-01	1.966e-02	5.969	2.43e-09 ***
zipcode98023	-4.360e-02	1.175e-02	-3.712	0.000207 ***
zipcode98024	3.976e-01	3.442e-02	11.549	< 2e-16 ***
zipcode98027	5.065e-01	2.173e-02	23.314	< 2e-16 ***
zipcode98028	2.608e-01	3.082e-02	8.463	< 2e-16 ***
zipcode98029	5.590e-01	2.486e-02	22.485	< 2e-16 ***
zipcode98030	4.477e-02	1.400e-02	3.198	0.001385 **
zipcode98031	5.616e-02	1.461e-02	3.843	0.000122 ***
zipcode98032	-2.093e-02	1.754e-02	-1.193	0.232747
zipcode98033	6.620e-01	2.664e-02	24.844	< 2e-16 ***
zipcode98034	4.059e-01	2.849e-02	14.248	< 2e-16 ***
zipcode98038	1.813e-01	1.668e-02	10.866	< 2e-16 ***
zipcode98039	1.279e+00	3.486e-02	36.700	< 2e-16 ***
zipcode98040	8.226e-01	2.063e-02	39.869	< 2e-16 ***
zipcode98042	7.332e-02	1.384e-02	5.296	1.20e-07 ***
zipcode98045	3.885e-01	3.181e-02	12.213	< 2e-16 ***
zipcode98052	5.266e-01	2.743e-02	19.197	< 2e-16 ***

```

## zipcode98053 5.019e-01 2.978e-02 16.855 < 2e-16 ***
## zipcode98055 1.085e-01 1.647e-02 6.590 4.49e-11 ***
## zipcode98056 2.775e-01 1.786e-02 15.540 < 2e-16 ***
## zipcode98058 1.278e-01 1.567e-02 8.156 3.67e-16 ***
## zipcode98059 2.960e-01 1.769e-02 16.736 < 2e-16 ***
## zipcode98065 3.972e-01 2.883e-02 13.778 < 2e-16 ***
## zipcode98070 2.318e-01 2.247e-02 10.316 < 2e-16 ***
## zipcode98072 3.476e-01 3.190e-02 10.895 < 2e-16 ***
## zipcode98074 4.878e-01 2.636e-02 18.505 < 2e-16 ***
## zipcode98075 5.036e-01 2.547e-02 19.777 < 2e-16 ***
## zipcode98077 3.393e-01 3.344e-02 10.146 < 2e-16 ***
## zipcode98092 2.638e-02 1.277e-02 2.066 0.038797 *
## zipcode98102 8.785e-01 2.767e-02 31.745 < 2e-16 ***
## zipcode98103 7.079e-01 2.546e-02 27.808 < 2e-16 ***
## zipcode98105 8.209e-01 2.620e-02 31.327 < 2e-16 ***
## zipcode98106 3.115e-01 1.890e-02 16.475 < 2e-16 ***
## zipcode98107 7.255e-01 2.621e-02 27.675 < 2e-16 ***
## zipcode98108 3.114e-01 2.112e-02 14.746 < 2e-16 ***
## zipcode98109 8.601e-01 2.765e-02 31.104 < 2e-16 ***
## zipcode98112 9.361e-01 2.424e-02 38.615 < 2e-16 ***
## zipcode98115 6.977e-01 2.592e-02 26.920 < 2e-16 ***
## zipcode98116 6.753e-01 2.105e-02 32.083 < 2e-16 ***
## zipcode98117 6.860e-01 2.621e-02 26.179 < 2e-16 ***
## zipcode98118 4.075e-01 1.845e-02 22.090 < 2e-16 ***
## zipcode98119 8.593e-01 2.579e-02 33.314 < 2e-16 ***
## zipcode98122 7.048e-01 2.285e-02 30.846 < 2e-16 ***
## zipcode98125 4.325e-01 2.802e-02 15.435 < 2e-16 ***
## zipcode98126 4.939e-01 1.936e-02 25.510 < 2e-16 ***
## zipcode98133 3.229e-01 2.886e-02 11.186 < 2e-16 ***
## zipcode98136 6.131e-01 1.987e-02 30.853 < 2e-16 ***
## zipcode98144 5.891e-01 2.123e-02 27.747 < 2e-16 ***
## zipcode98146 2.224e-01 1.793e-02 12.401 < 2e-16 ***
## zipcode98148 1.495e-01 2.771e-02 5.395 6.93e-08 ***
## zipcode98155 2.939e-01 3.008e-02 9.771 < 2e-16 ***
## zipcode98166 2.635e-01 1.634e-02 16.131 < 2e-16 ***
## zipcode98168 6.402e-02 1.716e-02 3.732 0.000191 ***
## zipcode98177 3.928e-01 3.029e-02 12.967 < 2e-16 ***
## zipcode98178 1.127e-01 1.779e-02 6.335 2.43e-10 ***
## zipcode98188 6.673e-02 1.855e-02 3.596 0.000324 ***
## zipcode98198 4.129e-02 1.383e-02 2.985 0.002844 **
## zipcode98199 7.254e-01 2.488e-02 29.155 < 2e-16 ***
## lat 3.630e-01 6.301e-02 5.760 8.52e-09 ***
## long -1.997e-01 5.095e-02 -3.919 8.92e-05 ***
## sqft_living15 6.622e-05 2.948e-06 22.465 < 2e-16 ***
## sqft_lot15 5.253e-08 7.546e-08 0.696 0.486318
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1531 on 20287 degrees of freedom
## Multiple R-squared: 0.8346, Adjusted R-squared: 0.8339
## F-statistic: 1191 on 86 and 20287 DF, p-value: < 2.2e-16

```

```
# Checking for multicollinearity
vif(lm3)
```

```
##              GVIF Df GVIF^(1/(2*Df))
## date          1.011040 1          1.005505
## bedrooms      1.771346 1          1.330919
## bathrooms     3.280819 1          1.811303
## sqft_lot       2.578861 1          1.605883
## floors         2.511135 1          1.584656
## waterfront     1.164729 1          1.079226
## view           1.429693 1          1.195698
## condition      1.347910 1          1.160995
## grade          3.682230 1          1.918914
## sqft_above     5.480691 1          2.341088
## sqft_basement  2.143457 1          1.464055
## yr_built       3.147902 1          1.774233
## yr_renovated   1.166822 1          1.080195
## zipcode       8923.672817 69        1.068137
## lat            67.094646 1          8.191132
## long           43.426209 1          6.589857
## sqft_living15  3.417277 1          1.848588
## sqft_lot15     2.725768 1          1.650990
```

```
# Dropping zipcode as it is colinear with Lat and Long
lm4 <- lm(log(price_sqft) ~ . -zipcode, data = data2)

summary(lm4)
```



```
##
## Call:
## lm(formula = log(price_sqft) ~ . - zipcode, data = data2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.89301 -0.16046 -0.00222  0.15000  0.99761
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -6.590e+01  1.979e+00 -33.298 < 2e-16 ***
## date          1.820e-04  1.470e-05  12.380 < 2e-16 ***
## bedrooms     -4.925e-02  2.459e-03 -20.024 < 2e-16 ***
## bathrooms     4.739e-02  4.042e-03  11.724 < 2e-16 ***
## sqft_lot      5.970e-07  8.278e-08   7.212 5.71e-13 ***
## floors        4.867e-02  4.375e-03  11.123 < 2e-16 ***
## waterfront1   4.447e-01  2.571e-02  17.298 < 2e-16 ***
## view          6.705e-02  2.661e-03  25.196 < 2e-16 ***
## condition     5.312e-02  2.885e-03  18.416 < 2e-16 ***
## grade         1.568e-01  2.673e-03  58.648 < 2e-16 ***
## sqft_above    -2.552e-04  4.749e-06 -53.742 < 2e-16 ***
## sqft_basement -2.880e-04  5.574e-06 -51.672 < 2e-16 ***
## yr_built      -3.736e-03  8.961e-05 -41.690 < 2e-16 ***
## yr_renovated   3.280e-05  4.567e-06   7.182 7.12e-13 ***
## lat           1.372e+00  1.263e-02 108.629 < 2e-16 ***
## long          -7.751e-02  1.481e-02  -5.232 1.69e-07 ***
## sqft_living15  8.202e-05  4.331e-06  18.937 < 2e-16 ***
## sqft_lot15    -3.004e-07  1.134e-07  -2.648  0.0081 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2363 on 20356 degrees of freedom
## Multiple R-squared:  0.6044, Adjusted R-squared:  0.6041
## F-statistic: 1829 on 17 and 20356 DF, p-value: < 2.2e-16
```

```
# Replacing Lat and Long with zipcode since the R2 value dropped drastically
lm5 <- lm(log(price_sqft) ~ . - lat - long, data=data2)
summary(lm5)
```

```
##
## Call:
## lm(formula = log(price_sqft) ~ . - lat - long, data = data2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.52812 -0.09746 -0.00036  0.09593  0.52124
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  2.571e+00  2.028e-01  12.678 < 2e-16 ***
## date         1.950e-04  9.553e-06  20.414 < 2e-16 ***
## bedrooms    -3.283e-02  1.622e-03 -20.242 < 2e-16 ***
## bathrooms    1.833e-02  2.646e-03   6.928 4.41e-12 ***
## sqft_lot     8.649e-07  5.388e-08  16.052 < 2e-16 ***
## floors      -6.217e-02  3.142e-03 -19.787 < 2e-16 ***
## waterfront1  5.450e-01  1.683e-02  32.389 < 2e-16 ***
## view         6.552e-02  1.778e-03  36.855 < 2e-16 ***
## condition    4.387e-02  1.931e-03  22.718 < 2e-16 ***
## grade        8.454e-02  1.835e-03  46.061 < 2e-16 ***
## sqft_above   -1.857e-04  3.177e-06 -58.439 < 2e-16 ***
## sqft_basement -3.154e-04  3.663e-06 -86.113 < 2e-16 ***
## yr_built     -5.737e-04  6.549e-05  -8.759 < 2e-16 ***
## yr_renovated  3.445e-05  2.989e-06  11.524 < 2e-16 ***
## zipcode98002  1.515e-03  1.390e-02   0.109  0.9132
## zipcode98003  2.098e-02  1.248e-02   1.680  0.0929 .
## zipcode98004  1.153e+00  1.241e-02  92.897 < 2e-16 ***
## zipcode98005  7.459e-01  1.488e-02  50.136 < 2e-16 ***
## zipcode98006  6.624e-01  1.106e-02  59.893 < 2e-16 ***
## zipcode98007  6.619e-01  1.562e-02  42.378 < 2e-16 ***
## zipcode98008  6.453e-01  1.246e-02  51.789 < 2e-16 ***
## zipcode98010  2.074e-01  1.967e-02  10.548 < 2e-16 ***
## zipcode98011  4.465e-01  1.387e-02  32.193 < 2e-16 ***
## zipcode98014  3.252e-01  1.913e-02  17.004 < 2e-16 ***
## zipcode98019  3.274e-01  1.429e-02  22.910 < 2e-16 ***
## zipcode98022  2.834e-02  1.339e-02   2.117  0.0343 *
## zipcode98023 -2.548e-02  1.076e-02  -2.369  0.0178 *
## zipcode98024  4.178e-01  2.364e-02  17.675 < 2e-16 ***
## zipcode98027  5.387e-01  1.143e-02  47.123 < 2e-16 ***
## zipcode98028  4.174e-01  1.241e-02  33.625 < 2e-16 ***
## zipcode98029  5.966e-01  1.201e-02  49.690 < 2e-16 ***
## zipcode98030  5.023e-02  1.275e-02   3.940 8.19e-05 ***
## zipcode98031  7.408e-02  1.250e-02   5.924 3.19e-09 ***
## zipcode98032  3.704e-03  1.711e-02   0.217  0.8286
## zipcode98033  7.796e-01  1.128e-02  69.122 < 2e-16 ***
## zipcode98034  5.423e-01  1.060e-02  51.157 < 2e-16 ***
## zipcode98038  1.559e-01  1.047e-02  14.897 < 2e-16 ***
## zipcode98039  1.386e+00  2.851e-02  48.624 < 2e-16 ***
## zipcode98040  9.042e-01  1.284e-02  70.436 < 2e-16 ***
## zipcode98042  6.298e-02  1.058e-02   5.951 2.70e-09 ***
## zipcode98045  3.473e-01  1.362e-02  25.499 < 2e-16 ***
## zipcode98052  6.308e-01  1.053e-02  59.928 < 2e-16 ***
```

```

## zipcode98053 5.880e-01 1.145e-02 51.374 < 2e-16 ***
## zipcode98055 1.483e-01 1.289e-02 11.506 < 2e-16 ***
## zipcode98056 3.325e-01 1.134e-02 29.312 < 2e-16 ***
## zipcode98058 1.527e-01 1.101e-02 13.868 < 2e-16 ***
## zipcode98059 3.365e-01 1.099e-02 30.608 < 2e-16 ***
## zipcode98065 3.967e-01 1.217e-02 32.605 < 2e-16 ***
## zipcode98070 3.113e-01 1.921e-02 16.206 < 2e-16 ***
## zipcode98072 4.806e-01 1.258e-02 38.203 < 2e-16 ***
## zipcode98074 5.566e-01 1.121e-02 49.657 < 2e-16 ***
## zipcode98075 5.549e-01 1.183e-02 46.905 < 2e-16 ***
## zipcode98077 4.569e-01 1.413e-02 32.336 < 2e-16 ***
## zipcode98092 4.155e-03 1.177e-02 0.353 0.7240
## zipcode98102 1.006e+00 1.868e-02 53.850 < 2e-16 ***
## zipcode98103 8.551e-01 1.080e-02 79.177 < 2e-16 ***
## zipcode98105 9.542e-01 1.369e-02 69.695 < 2e-16 ***
## zipcode98106 4.107e-01 1.211e-02 33.917 < 2e-16 ***
## zipcode98107 8.766e-01 1.294e-02 67.759 < 2e-16 ***
## zipcode98108 4.056e-01 1.474e-02 27.521 < 2e-16 ***
## zipcode98109 9.941e-01 1.843e-02 53.952 < 2e-16 ***
## zipcode98112 1.057e+00 1.357e-02 77.884 < 2e-16 ***
## zipcode98115 8.397e-01 1.072e-02 78.323 < 2e-16 ***
## zipcode98116 7.952e-01 1.219e-02 65.254 < 2e-16 ***
## zipcode98117 8.440e-01 1.083e-02 77.911 < 2e-16 ***
## zipcode98118 4.931e-01 1.107e-02 44.532 < 2e-16 ***
## zipcode98119 9.977e-01 1.511e-02 66.041 < 2e-16 ***
## zipcode98122 8.190e-01 1.290e-02 63.462 < 2e-16 ***
## zipcode98125 5.861e-01 1.142e-02 51.338 < 2e-16 ***
## zipcode98126 5.999e-01 1.195e-02 50.215 < 2e-16 ***
## zipcode98133 4.918e-01 1.088e-02 45.192 < 2e-16 ***
## zipcode98136 7.193e-01 1.299e-02 55.362 < 2e-16 ***
## zipcode98144 6.947e-01 1.217e-02 57.062 < 2e-16 ***
## zipcode98146 3.096e-01 1.288e-02 24.041 < 2e-16 ***
## zipcode98148 2.061e-01 2.654e-02 7.767 8.39e-15 ***
## zipcode98155 4.621e-01 1.109e-02 41.660 < 2e-16 ***
## zipcode98166 3.320e-01 1.321e-02 25.142 < 2e-16 ***
## zipcode98168 1.370e-01 1.288e-02 10.635 < 2e-16 ***
## zipcode98177 5.697e-01 1.313e-02 43.375 < 2e-16 ***
## zipcode98178 1.765e-01 1.301e-02 13.569 < 2e-16 ***
## zipcode98188 1.188e-01 1.642e-02 7.239 4.67e-13 ***
## zipcode98198 7.981e-02 1.274e-02 6.263 3.86e-10 ***
## zipcode98199 8.730e-01 1.231e-02 70.909 < 2e-16 ***
## sqft_living15 6.641e-05 2.951e-06 22.504 < 2e-16 ***
## sqft_lot15 2.940e-08 7.531e-08 0.390 0.6963
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1533 on 20289 degrees of freedom
## Multiple R-squared:  0.8342, Adjusted R-squared:  0.8335
## F-statistic: 1215 on 84 and 20289 DF, p-value: < 2.2e-16

```

```
data3 <- data2[,c(-15, -16, -18)]  
  
lm6 <- lm(log(price_sqft)~ ., data = data3)  
  
summary(lm6)
```

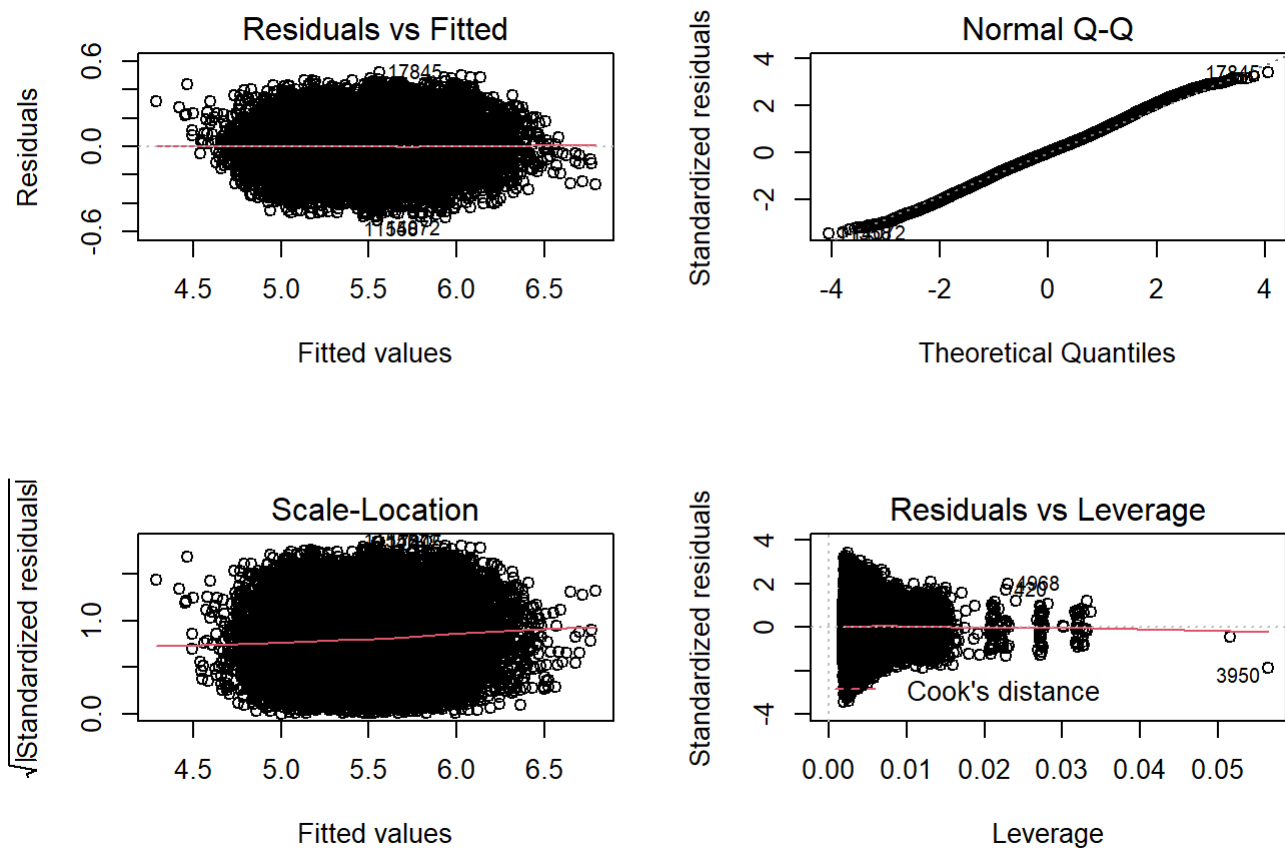
```
##
## Call:
## lm(formula = log(price_sqft) ~ ., data = data3)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.52816 -0.09740 -0.00042  0.09591  0.52129
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  2.571e+00  2.028e-01  12.679 < 2e-16 ***
## date         1.950e-04  9.553e-06  20.414 < 2e-16 ***
## bedrooms    -3.284e-02  1.622e-03 -20.254 < 2e-16 ***
## bathrooms    1.831e-02  2.646e-03   6.922 4.60e-12 ***
## sqft_lot     8.801e-07  3.720e-08  23.659 < 2e-16 ***
## floors      -6.220e-02  3.141e-03 -19.801 < 2e-16 ***
## waterfront1  5.450e-01  1.683e-02  32.393 < 2e-16 ***
## view         6.552e-02  1.778e-03  36.855 < 2e-16 ***
## condition    4.387e-02  1.931e-03  22.724 < 2e-16 ***
## grade        8.453e-02  1.835e-03  46.060 < 2e-16 ***
## sqft_above   -1.856e-04  3.177e-06 -58.442 < 2e-16 ***
## sqft_basement -3.154e-04  3.662e-06 -86.118 < 2e-16 ***
## yr_built     -5.737e-04  6.549e-05  -8.760 < 2e-16 ***
## yr_renovated  3.445e-05  2.989e-06  11.527 < 2e-16 ***
## zipcode98002  1.529e-03  1.390e-02   0.110  0.9124
## zipcode98003  2.099e-02  1.248e-02   1.681  0.0928 .
## zipcode98004  1.153e+00  1.241e-02  92.900 < 2e-16 ***
## zipcode98005  7.460e-01  1.488e-02  50.145 < 2e-16 ***
## zipcode98006  6.624e-01  1.106e-02  59.893 < 2e-16 ***
## zipcode98007  6.618e-01  1.562e-02  42.378 < 2e-16 ***
## zipcode98008  6.453e-01  1.246e-02  51.789 < 2e-16 ***
## zipcode98010  2.080e-01  1.962e-02  10.599 < 2e-16 ***
## zipcode98011  4.465e-01  1.387e-02  32.192 < 2e-16 ***
## zipcode98014  3.257e-01  1.910e-02  17.055 < 2e-16 ***
## zipcode98019  3.278e-01  1.427e-02  22.977 < 2e-16 ***
## zipcode98022  2.871e-02  1.336e-02   2.149  0.0316 *
## zipcode98023 -2.548e-02  1.076e-02  -2.369  0.0179 *
## zipcode98024  4.186e-01  2.355e-02  17.777 < 2e-16 ***
## zipcode98027  5.390e-01  1.142e-02  47.196 < 2e-16 ***
## zipcode98028  4.174e-01  1.241e-02  33.624 < 2e-16 ***
## zipcode98029  5.966e-01  1.201e-02  49.689 < 2e-16 ***
## zipcode98030  5.019e-02  1.275e-02   3.936 8.30e-05 ***
## zipcode98031  7.403e-02  1.250e-02   5.921 3.25e-09 ***
## zipcode98032  3.706e-03  1.711e-02   0.217  0.8285
## zipcode98033  7.796e-01  1.128e-02  69.123 < 2e-16 ***
## zipcode98034  5.423e-01  1.060e-02  51.157 < 2e-16 ***
## zipcode98038  1.560e-01  1.047e-02  14.906 < 2e-16 ***
## zipcode98039  1.386e+00  2.851e-02  48.626 < 2e-16 ***
## zipcode98040  9.042e-01  1.284e-02  70.437 < 2e-16 ***
## zipcode98042  6.300e-02  1.058e-02   5.954 2.67e-09 ***
## zipcode98045  3.475e-01  1.361e-02  25.524 < 2e-16 ***
## zipcode98052  6.308e-01  1.053e-02  59.929 < 2e-16 ***
```

```

## zipcode98053 5.881e-01 1.144e-02 51.418 < 2e-16 ***
## zipcode98055 1.483e-01 1.289e-02 11.504 < 2e-16 ***
## zipcode98056 3.325e-01 1.134e-02 29.310 < 2e-16 ***
## zipcode98058 1.527e-01 1.101e-02 13.867 < 2e-16 ***
## zipcode98059 3.365e-01 1.099e-02 30.609 < 2e-16 ***
## zipcode98065 3.966e-01 1.216e-02 32.604 < 2e-16 ***
## zipcode98070 3.120e-01 1.912e-02 16.316 < 2e-16 ***
## zipcode98072 4.807e-01 1.258e-02 38.215 < 2e-16 ***
## zipcode98074 5.566e-01 1.121e-02 49.659 < 2e-16 ***
## zipcode98075 5.549e-01 1.183e-02 46.905 < 2e-16 ***
## zipcode98077 4.572e-01 1.410e-02 32.420 < 2e-16 ***
## zipcode98092 4.392e-03 1.175e-02 0.374 0.7086
## zipcode98102 1.006e+00 1.868e-02 53.850 < 2e-16 ***
## zipcode98103 8.550e-01 1.080e-02 79.178 < 2e-16 ***
## zipcode98105 9.541e-01 1.369e-02 69.696 < 2e-16 ***
## zipcode98106 4.107e-01 1.211e-02 33.917 < 2e-16 ***
## zipcode98107 8.765e-01 1.294e-02 67.759 < 2e-16 ***
## zipcode98108 4.056e-01 1.474e-02 27.520 < 2e-16 ***
## zipcode98109 9.940e-01 1.842e-02 53.951 < 2e-16 ***
## zipcode98112 1.057e+00 1.357e-02 77.886 < 2e-16 ***
## zipcode98115 8.397e-01 1.072e-02 78.324 < 2e-16 ***
## zipcode98116 7.952e-01 1.219e-02 65.254 < 2e-16 ***
## zipcode98117 8.439e-01 1.083e-02 77.912 < 2e-16 ***
## zipcode98118 4.931e-01 1.107e-02 44.531 < 2e-16 ***
## zipcode98119 9.977e-01 1.511e-02 66.042 < 2e-16 ***
## zipcode98122 8.189e-01 1.290e-02 63.462 < 2e-16 ***
## zipcode98125 5.861e-01 1.142e-02 51.338 < 2e-16 ***
## zipcode98126 5.999e-01 1.195e-02 50.215 < 2e-16 ***
## zipcode98133 4.918e-01 1.088e-02 45.193 < 2e-16 ***
## zipcode98136 7.193e-01 1.299e-02 55.362 < 2e-16 ***
## zipcode98144 6.947e-01 1.217e-02 57.063 < 2e-16 ***
## zipcode98146 3.096e-01 1.288e-02 24.042 < 2e-16 ***
## zipcode98148 2.061e-01 2.654e-02 7.769 8.31e-15 ***
## zipcode98155 4.621e-01 1.109e-02 41.663 < 2e-16 ***
## zipcode98166 3.321e-01 1.321e-02 25.145 < 2e-16 ***
## zipcode98168 1.370e-01 1.288e-02 10.636 < 2e-16 ***
## zipcode98177 5.697e-01 1.313e-02 43.377 < 2e-16 ***
## zipcode98178 1.765e-01 1.301e-02 13.569 < 2e-16 ***
## zipcode98188 1.189e-01 1.641e-02 7.241 4.60e-13 ***
## zipcode98198 7.982e-02 1.274e-02 6.263 3.85e-10 ***
## zipcode98199 8.729e-01 1.231e-02 70.911 < 2e-16 ***
## sqft_living15 6.649e-05 2.944e-06 22.587 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1533 on 20290 degrees of freedom
## Multiple R-squared: 0.8342, Adjusted R-squared: 0.8335
## F-statistic: 1230 on 83 and 20290 DF, p-value: < 2.2e-16

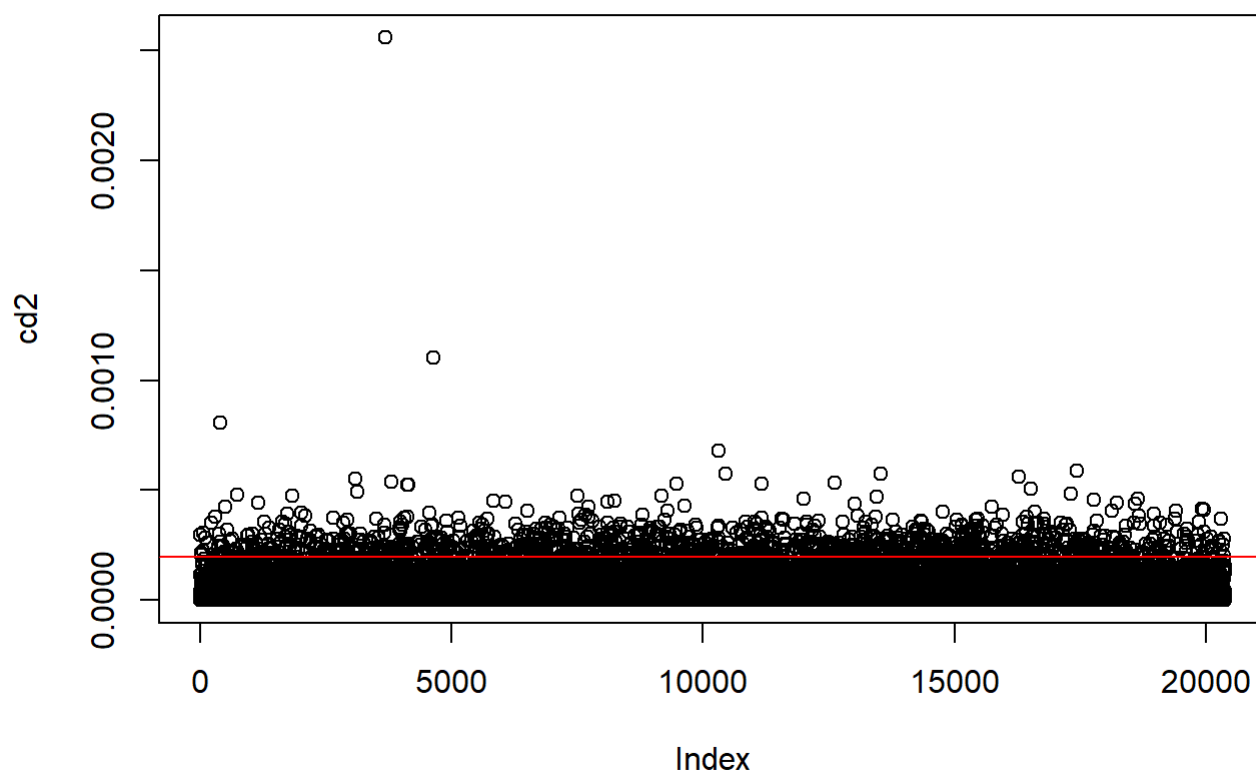
```

```
# Checking the assumptions of linear models
par(mfrow=c(2,2))
plot(lm6)
```



```
cd2 <- cooks.distance(lm6)

plot(cd2)
abline(h = 4/nrow(data3), col='red')
```



```
data4 <- data2[-which(cd > 4/nrow(data3)),]  
  
lm7 <- lm(log(price_sqft) ~ ., data = data4)  
summary(lm7)
```



```
##
## Call:
## lm(formula = log(price_sqft) ~ ., data = data4)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.53581 -0.09798 -0.00043  0.09616  0.52106
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -4.126e+01  6.959e+00  -5.929 3.10e-09 ***
## date          1.918e-04  9.800e-06  19.567 < 2e-16 ***
## bedrooms     -3.277e-02  1.663e-03 -19.705 < 2e-16 ***
## bathrooms     1.885e-02  2.717e-03   6.938 4.12e-12 ***
## sqft_lot      8.754e-07  5.486e-08  15.956 < 2e-16 ***
## floors       -6.395e-02  3.235e-03 -19.770 < 2e-16 ***
## waterfront1   5.461e-01  1.717e-02  31.807 < 2e-16 ***
## view          6.551e-02  1.823e-03  35.927 < 2e-16 ***
## condition     4.374e-02  1.981e-03  22.073 < 2e-16 ***
## grade         8.443e-02  1.884e-03  44.820 < 2e-16 ***
## sqft_above    -1.848e-04  3.269e-06 -56.545 < 2e-16 ***
## sqft_basement -3.167e-04  3.768e-06 -84.051 < 2e-16 ***
## yr_built      -5.594e-04  6.728e-05  -8.315 < 2e-16 ***
## yr_renovated   3.393e-05  3.064e-06  11.074 < 2e-16 ***
## zipcode98002   1.598e-02  1.474e-02   1.084 0.278320
## zipcode98003   9.759e-03  1.286e-02   0.759 0.447783
## zipcode98004   1.050e+00  2.405e-02  43.650 < 2e-16 ***
## zipcode98005   6.557e-01  2.578e-02  25.436 < 2e-16 ***
## zipcode98006   5.985e-01  2.133e-02  28.055 < 2e-16 ***
## zipcode98007   5.774e-01  2.672e-02  21.606 < 2e-16 ***
## zipcode98008   5.654e-01  2.563e-02  22.055 < 2e-16 ***
## zipcode98010   2.567e-01  2.485e-02  10.332 < 2e-16 ***
## zipcode98011   3.016e-01  3.272e-02   9.219 < 2e-16 ***
## zipcode98014   2.820e-01  3.803e-02   7.415 1.27e-13 ***
## zipcode98019   2.380e-01  3.678e-02   6.470 1.00e-10 ***
## zipcode98022   1.229e-01  2.033e-02   6.049 1.49e-09 ***
## zipcode98023  -4.506e-02  1.212e-02  -3.718 0.000201 ***
## zipcode98024   3.995e-01  3.574e-02  11.178 < 2e-16 ***
## zipcode98027   5.093e-01  2.253e-02  22.605 < 2e-16 ***
## zipcode98028   2.549e-01  3.164e-02   8.057 8.24e-16 ***
## zipcode98029   5.639e-01  2.586e-02  21.804 < 2e-16 ***
## zipcode98030   4.530e-02  1.441e-02   3.145 0.001663 **
## zipcode98031   5.710e-02  1.500e-02   3.808 0.000140 ***
## zipcode98032  -1.876e-02  1.801e-02  -1.042 0.297460
## zipcode98033   6.581e-01  2.740e-02  24.019 < 2e-16 ***
## zipcode98034   4.032e-01  2.927e-02  13.776 < 2e-16 ***
## zipcode98038   1.837e-01  1.739e-02  10.568 < 2e-16 ***
## zipcode98039   1.284e+00  3.626e-02  35.399 < 2e-16 ***
## zipcode98040   8.228e-01  2.123e-02  38.751 < 2e-16 ***
## zipcode98042   7.434e-02  1.437e-02   5.174 2.32e-07 ***
## zipcode98045   3.913e-01  3.340e-02  11.718 < 2e-16 ***
## zipcode98052   5.242e-01  2.826e-02  18.549 < 2e-16 ***
```

```

## zipcode98053 5.035e-01 3.082e-02 16.339 < 2e-16 ***
## zipcode98055 1.060e-01 1.693e-02 6.260 3.93e-10 ***
## zipcode98056 2.799e-01 1.840e-02 15.210 < 2e-16 ***
## zipcode98058 1.283e-01 1.616e-02 7.936 2.20e-15 ***
## zipcode98059 2.964e-01 1.829e-02 16.206 < 2e-16 ***
## zipcode98065 4.019e-01 3.009e-02 13.358 < 2e-16 ***
## zipcode98070 2.304e-01 2.330e-02 9.889 < 2e-16 ***
## zipcode98072 3.437e-01 3.281e-02 10.474 < 2e-16 ***
## zipcode98074 4.884e-01 2.726e-02 17.919 < 2e-16 ***
## zipcode98075 5.050e-01 2.636e-02 19.155 < 2e-16 ***
## zipcode98077 3.340e-01 3.459e-02 9.656 < 2e-16 ***
## zipcode98092 2.635e-02 1.315e-02 2.004 0.045094 *
## zipcode98102 8.777e-01 2.825e-02 31.071 < 2e-16 ***
## zipcode98103 7.068e-01 2.610e-02 27.081 < 2e-16 ***
## zipcode98105 8.145e-01 2.687e-02 30.311 < 2e-16 ***
## zipcode98106 3.052e-01 1.937e-02 15.757 < 2e-16 ***
## zipcode98107 7.240e-01 2.687e-02 26.946 < 2e-16 ***
## zipcode98108 3.093e-01 2.160e-02 14.316 < 2e-16 ***
## zipcode98109 8.514e-01 2.832e-02 30.068 < 2e-16 ***
## zipcode98112 9.332e-01 2.486e-02 37.538 < 2e-16 ***
## zipcode98115 6.951e-01 2.658e-02 26.148 < 2e-16 ***
## zipcode98116 6.732e-01 2.162e-02 31.144 < 2e-16 ***
## zipcode98117 6.809e-01 2.688e-02 25.337 < 2e-16 ***
## zipcode98118 4.044e-01 1.892e-02 21.381 < 2e-16 ***
## zipcode98119 8.545e-01 2.648e-02 32.264 < 2e-16 ***
## zipcode98122 7.058e-01 2.345e-02 30.093 < 2e-16 ***
## zipcode98125 4.295e-01 2.870e-02 14.965 < 2e-16 ***
## zipcode98126 4.920e-01 1.985e-02 24.790 < 2e-16 ***
## zipcode98133 3.172e-01 2.959e-02 10.718 < 2e-16 ***
## zipcode98136 6.101e-01 2.040e-02 29.907 < 2e-16 ***
## zipcode98144 5.906e-01 2.178e-02 27.119 < 2e-16 ***
## zipcode98146 2.237e-01 1.841e-02 12.149 < 2e-16 ***
## zipcode98148 1.474e-01 2.780e-02 5.301 1.16e-07 ***
## zipcode98155 2.890e-01 3.086e-02 9.364 < 2e-16 ***
## zipcode98166 2.622e-01 1.679e-02 15.619 < 2e-16 ***
## zipcode98168 5.788e-02 1.763e-02 3.283 0.001027 **
## zipcode98177 3.841e-01 3.106e-02 12.367 < 2e-16 ***
## zipcode98178 1.106e-01 1.823e-02 6.070 1.30e-09 ***
## zipcode98188 6.258e-02 1.897e-02 3.298 0.000974 ***
## zipcode98198 4.229e-02 1.425e-02 2.968 0.003004 **
## zipcode98199 7.218e-01 2.550e-02 28.310 < 2e-16 ***
## lat 3.723e-01 6.467e-02 5.756 8.73e-09 ***
## long -2.146e-01 5.380e-02 -3.990 6.64e-05 ***
## sqft_living15 6.621e-05 3.035e-06 21.812 < 2e-16 ***
## sqft_lot15 2.967e-08 7.751e-08 0.383 0.701909
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1529 on 19179 degrees of freedom
## Multiple R-squared: 0.8353, Adjusted R-squared: 0.8346
## F-statistic: 1131 on 86 and 19179 DF, p-value: < 2.2e-16

```

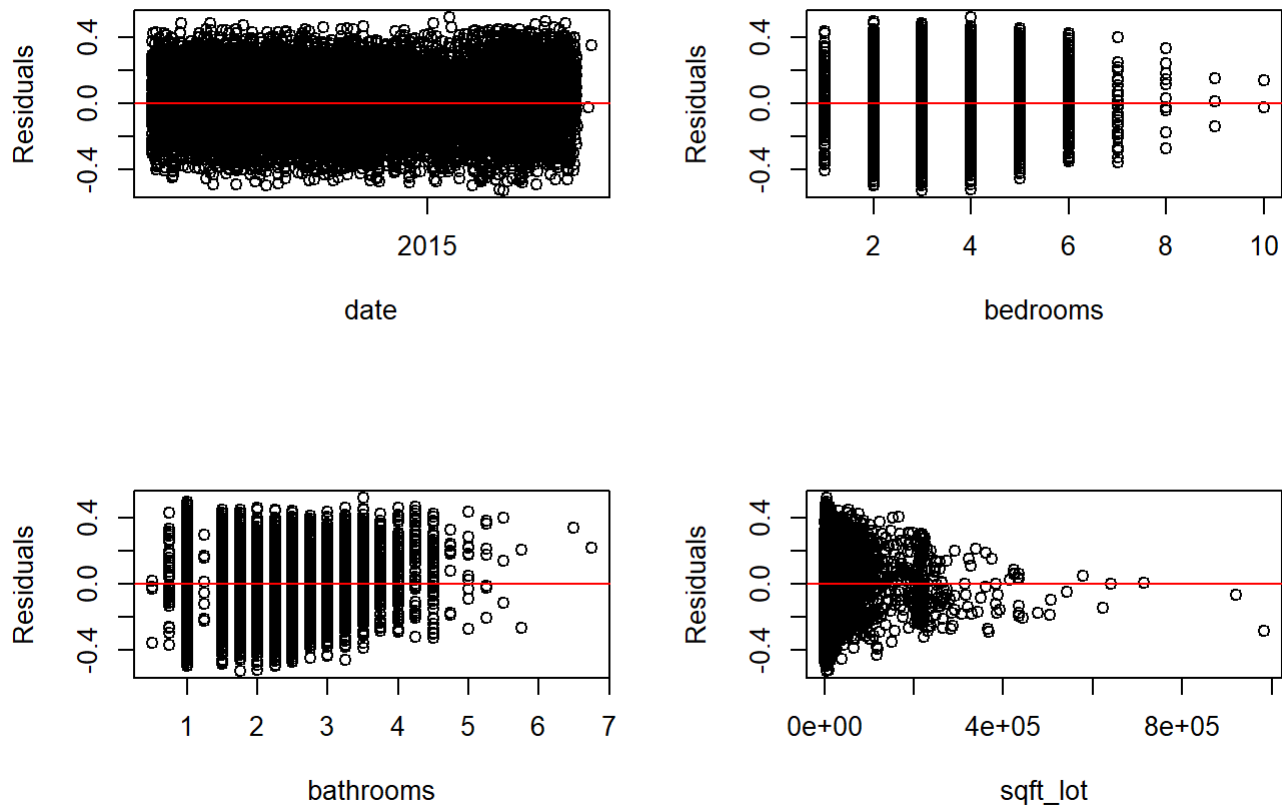
```

num.cols2 <- colnames(data3[,c(-6, -14, -16)])

par(mfrow = c(2,2))

for (i in num.cols2[1:4]) {
  plot(data3[,i], lm6$residuals, xlab = i, ylab = "Residuals")
  abline(h = 0, col='red')
}

```

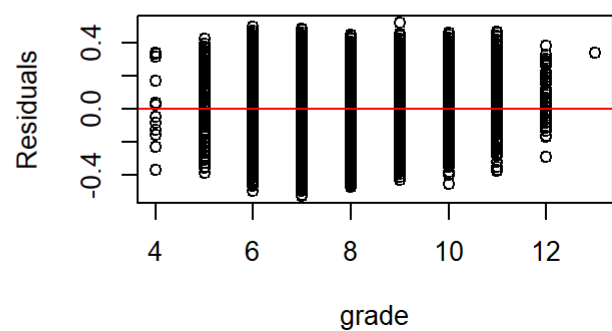
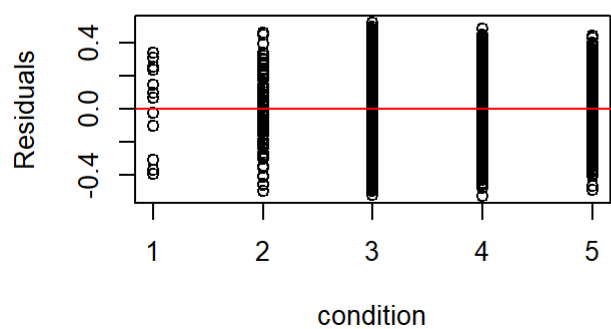
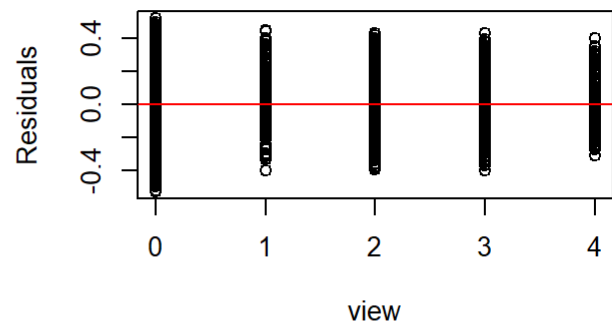
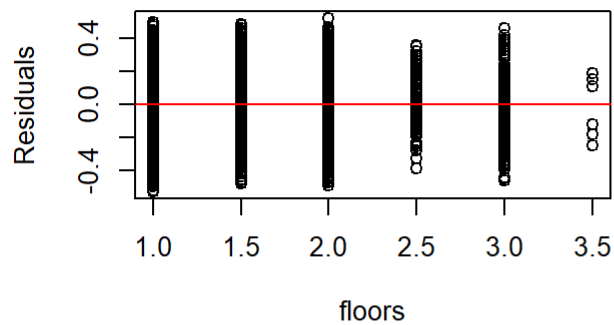


```

par(mfrow = c(2,2))

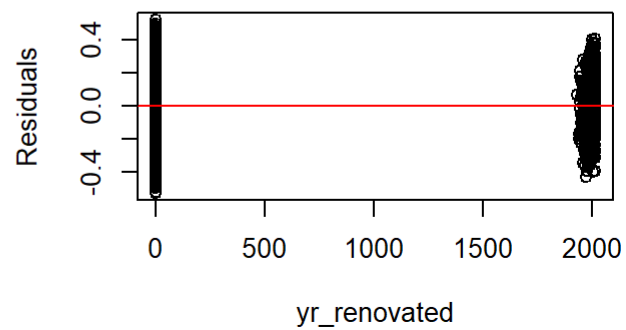
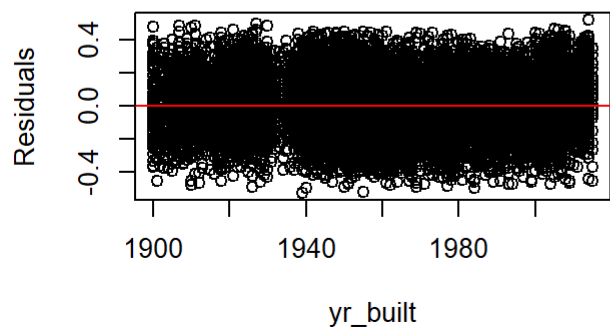
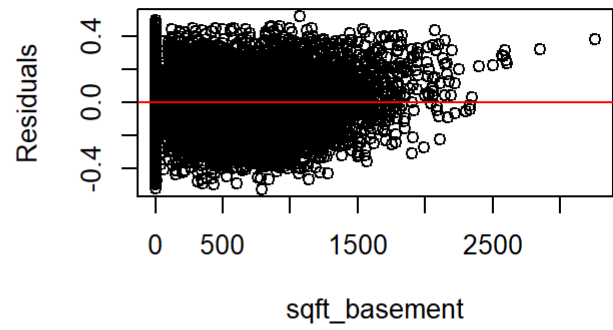
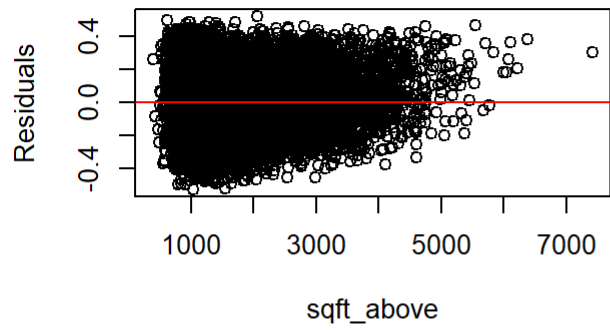
for (i in num.cols2[5:8]) {
  plot(data3[,i], lm6$residuals, xlab = i, ylab = "Residuals")
  abline(h = 0, col='red')
}

```



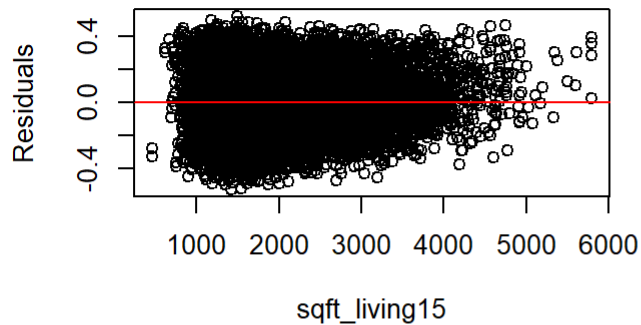
```
par(mfrow = c(2,2))

for (i in num.cols2[9:12]) {
  plot(data3[,i], lm6$residuals, xlab = i, ylab = "Residuals")
  abline(h = 0, col='red')
}
```



```
par(mfrow = c(2,2))

for (i in num.cols2[13]) {
  plot(data3[,i], lm6$residuals, xlab = i, ylab = "Residuals")
  abline(h = 0, col='red')
}
```



```
# Random Forest Regression to determine feature importance
```

```
rf <- randomForest(log(price_sqft) ~ ., data = data2[,c(-6,-14)] +
                    as.numeric(data2$zipcode) + as.numeric(data2$waterfront),
                    importance = TRUE)
```

```
rf
```

```
##
## Call:
## randomForest(formula = log(price_sqft) ~ ., data = data2[, c(-6, -14)] + as.numeric(data2$zipcode) + as.numeric(data2$waterfront), importance = TRUE)
##           Type of random forest: regression
##           Number of trees: 500
## No. of variables tried at each split: 5
##
##           Mean of squared residuals: 0.01780617
##           % Var explained: 85.75
```

```
rf$importance
```

```
##                %IncMSE IncNodePurity
## date           0.0006865678      41.60164
## bedrooms       0.0276198014     114.98868
## bathrooms      0.0253466280     106.59856
## sqft_lot       0.0066962638      63.94748
## floors         0.0386040547     143.06593
## view          0.1131407218     287.78255
## condition      0.0197439512      81.70609
## grade          0.0162194001      46.87869
## sqft_above     0.0120594979      94.81218
## sqft_basement  0.0203245059     135.24056
## yr_built       0.0114057350      70.13222
## yr_renovated   0.0809705164     356.69799
## lat            0.1126746622     402.55743
## long           0.1147148846     406.32394
## sqft_living15  0.0069931417      72.05548
## sqft_lot15     0.0103657598      89.32404
```

```
# Predictive models
```

```
# Creating a test and training set
```

```
set.seed(1)
```

```
flag <- sample(nrow(data2), nrow(data2)/10, replace = FALSE)
```

```
train <- data2[-flag,]
```

```
test <- data2[flag,]
```

```
lm6.p <- lm(log(price_sqft)~ . -lat -long -sqft_lot15, data = train)
```

```
summary(lm6.p)
```

```
##
## Call:
## lm(formula = log(price_sqft) ~ . - lat - long - sqft_lot15, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.52300 -0.09728  0.00010  0.09639  0.52842
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   2.536e+00  2.134e-01  11.882 < 2e-16 ***
## date          1.995e-04  1.005e-05  19.846 < 2e-16 ***
## bedrooms     -3.296e-02  1.704e-03 -19.348 < 2e-16 ***
## bathrooms     1.746e-02  2.780e-03   6.281 3.44e-10 ***
## sqft_lot      8.863e-07  4.006e-08  22.121 < 2e-16 ***
## floors       -5.995e-02  3.300e-03 -18.168 < 2e-16 ***
## waterfront1   5.342e-01  1.801e-02  29.652 < 2e-16 ***
## view          6.545e-02  1.876e-03  34.889 < 2e-16 ***
## condition     4.431e-02  2.033e-03  21.799 < 2e-16 ***
## grade         8.490e-02  1.934e-03  43.904 < 2e-16 ***
## sqft_above    -1.871e-04  3.355e-06 -55.777 < 2e-16 ***
## sqft_basement -3.142e-04  3.856e-06 -81.469 < 2e-16 ***
## yr_built      -5.943e-04  6.871e-05  -8.649 < 2e-16 ***
## yr_renovated   3.513e-05  3.135e-06  11.205 < 2e-16 ***
## zipcode98002  -2.019e-03  1.447e-02  -0.140 0.889013
## zipcode98003   1.548e-02  1.310e-02   1.182 0.237161
## zipcode98004   1.152e+00  1.292e-02  89.149 < 2e-16 ***
## zipcode98005   7.449e-01  1.536e-02  48.489 < 2e-16 ***
## zipcode98006   6.636e-01  1.158e-02  57.324 < 2e-16 ***
## zipcode98007   6.643e-01  1.646e-02  40.365 < 2e-16 ***
## zipcode98008   6.465e-01  1.311e-02  49.327 < 2e-16 ***
## zipcode98010   2.034e-01  2.056e-02   9.893 < 2e-16 ***
## zipcode98011   4.510e-01  1.468e-02  30.715 < 2e-16 ***
## zipcode98014   3.233e-01  1.993e-02  16.222 < 2e-16 ***
## zipcode98019   3.238e-01  1.482e-02  21.849 < 2e-16 ***
## zipcode98022   2.743e-02  1.381e-02   1.986 0.046999 *
## zipcode98023  -2.817e-02  1.132e-02  -2.489 0.012820 *
## zipcode98024   4.172e-01  2.430e-02  17.168 < 2e-16 ***
## zipcode98027   5.366e-01  1.200e-02  44.701 < 2e-16 ***
## zipcode98028   4.183e-01  1.294e-02  32.334 < 2e-16 ***
## zipcode98029   5.956e-01  1.260e-02  47.291 < 2e-16 ***
## zipcode98030   5.108e-02  1.338e-02   3.817 0.000135 ***
## zipcode98031   6.985e-02  1.312e-02   5.324 1.03e-07 ***
## zipcode98032   8.146e-04  1.785e-02   0.046 0.963597
## zipcode98033   7.779e-01  1.183e-02  65.774 < 2e-16 ***
## zipcode98034   5.397e-01  1.104e-02  48.892 < 2e-16 ***
## zipcode98038   1.537e-01  1.094e-02  14.044 < 2e-16 ***
## zipcode98039   1.385e+00  3.037e-02  45.613 < 2e-16 ***
## zipcode98040   9.067e-01  1.353e-02  67.029 < 2e-16 ***
## zipcode98042   6.163e-02  1.114e-02   5.534 3.18e-08 ***
## zipcode98045   3.472e-01  1.437e-02  24.161 < 2e-16 ***
## zipcode98052   6.271e-01  1.103e-02  56.836 < 2e-16 ***
```



```

## zipcode98053 5.882e-01 1.199e-02 49.038 < 2e-16 ***
## zipcode98055 1.447e-01 1.350e-02 10.722 < 2e-16 ***
## zipcode98056 3.270e-01 1.190e-02 27.472 < 2e-16 ***
## zipcode98058 1.552e-01 1.155e-02 13.437 < 2e-16 ***
## zipcode98059 3.365e-01 1.151e-02 29.248 < 2e-16 ***
## zipcode98065 3.977e-01 1.262e-02 31.523 < 2e-16 ***
## zipcode98070 3.128e-01 2.030e-02 15.404 < 2e-16 ***
## zipcode98072 4.818e-01 1.313e-02 36.706 < 2e-16 ***
## zipcode98074 5.506e-01 1.174e-02 46.917 < 2e-16 ***
## zipcode98075 5.520e-01 1.247e-02 44.258 < 2e-16 ***
## zipcode98077 4.542e-01 1.463e-02 31.045 < 2e-16 ***
## zipcode98092 1.960e-03 1.222e-02 0.160 0.872543
## zipcode98102 1.003e+00 1.973e-02 50.834 < 2e-16 ***
## zipcode98103 8.458e-01 1.131e-02 74.779 < 2e-16 ***
## zipcode98105 9.556e-01 1.437e-02 66.495 < 2e-16 ***
## zipcode98106 4.185e-01 1.272e-02 32.902 < 2e-16 ***
## zipcode98107 8.711e-01 1.362e-02 63.959 < 2e-16 ***
## zipcode98108 4.048e-01 1.535e-02 26.372 < 2e-16 ***
## zipcode98109 9.881e-01 1.954e-02 50.580 < 2e-16 ***
## zipcode98112 1.055e+00 1.407e-02 74.942 < 2e-16 ***
## zipcode98115 8.339e-01 1.117e-02 74.639 < 2e-16 ***
## zipcode98116 7.906e-01 1.276e-02 61.945 < 2e-16 ***
## zipcode98117 8.398e-01 1.134e-02 74.058 < 2e-16 ***
## zipcode98118 4.969e-01 1.160e-02 42.844 < 2e-16 ***
## zipcode98119 9.911e-01 1.593e-02 62.222 < 2e-16 ***
## zipcode98122 8.162e-01 1.342e-02 60.819 < 2e-16 ***
## zipcode98125 5.809e-01 1.187e-02 48.943 < 2e-16 ***
## zipcode98126 5.945e-01 1.250e-02 47.570 < 2e-16 ***
## zipcode98133 4.902e-01 1.140e-02 43.008 < 2e-16 ***
## zipcode98136 7.145e-01 1.359e-02 52.585 < 2e-16 ***
## zipcode98144 6.926e-01 1.270e-02 54.548 < 2e-16 ***
## zipcode98146 3.043e-01 1.338e-02 22.742 < 2e-16 ***
## zipcode98148 2.046e-01 2.661e-02 7.689 1.56e-14 ***
## zipcode98155 4.570e-01 1.157e-02 39.487 < 2e-16 ***
## zipcode98166 3.327e-01 1.394e-02 23.867 < 2e-16 ***
## zipcode98168 1.370e-01 1.345e-02 10.185 < 2e-16 ***
## zipcode98177 5.669e-01 1.387e-02 40.864 < 2e-16 ***
## zipcode98178 1.762e-01 1.369e-02 12.871 < 2e-16 ***
## zipcode98188 1.256e-01 1.718e-02 7.309 2.80e-13 ***
## zipcode98198 7.552e-02 1.351e-02 5.591 2.30e-08 ***
## zipcode98199 8.754e-01 1.297e-02 67.503 < 2e-16 ***
## sqft_living15 6.661e-05 3.097e-06 21.510 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.153 on 18253 degrees of freedom
## Multiple R-squared:  0.8344, Adjusted R-squared:  0.8337
## F-statistic: 1108 on 83 and 18253 DF, p-value: < 2.2e-16

```

```
pred.lm6 <- exp(predict(lm6.p, test))  
lm6.mse <- mean((test$price_sqft - pred.lm6)^2)  
  
lm6.mse
```

```
## [1] 2061.513
```

```
step.p <- step(lm(log(price_sqft)~., data = train), direction = "both")
```

```

## Start:  AIC=-68804.47
## log(price_sqft) ~ date + bedrooms + bathrooms + sqft_lot + floors +
##   waterfront + view + condition + grade + sqft_above + sqft_basement +
##   yr_built + yr_renovated + zipcode + lat + long + sqft_living15 +
##   sqft_lot15
##
##           Df Sum of Sq    RSS    AIC
## - sqft_lot15      1      0.01  426.24 -68806
## <none>                        426.22 -68804
## - long            1      0.24  426.47 -68796
## - lat             1      0.75  426.97 -68774
## - bathrooms       1      0.93  427.15 -68767
## - yr_built         1      1.71  427.93 -68733
## - yr_renovated     1      2.94  429.16 -68681
## - sqft_lot         1      5.16  431.39 -68586
## - floors           1      7.67  433.90 -68479
## - bedrooms        1      8.71  434.94 -68435
## - date             1      9.16  435.39 -68417
## - sqft_living15    1     10.62  436.85 -68355
## - condition        1     11.27  437.49 -68328
## - waterfront       1     20.71  446.93 -67936
## - view             1     28.56  454.78 -67617
## - grade            1     44.61  470.84 -66981
## - sqft_above       1     72.29  498.51 -65934
## - sqft_basement    1    155.48  581.70 -63104
## - zipcode          69    595.67 1021.89 -52908
##
## Step:  AIC=-68805.88
## log(price_sqft) ~ date + bedrooms + bathrooms + sqft_lot + floors +
##   waterfront + view + condition + grade + sqft_above + sqft_basement +
##   yr_built + yr_renovated + zipcode + lat + long + sqft_living15
##
##           Df Sum of Sq    RSS    AIC
## <none>                        426.24 -68806
## + sqft_lot15      1      0.01  426.22 -68804
## - long            1      0.23  426.47 -68798
## - lat             1      0.75  426.98 -68776
## - bathrooms       1      0.92  427.16 -68768
## - yr_built         1      1.71  427.95 -68734
## - yr_renovated     1      2.94  429.18 -68682
## - floors           1      7.69  433.92 -68480
## - bedrooms        1      8.73  434.97 -68436
## - date             1      9.16  435.40 -68418
## - sqft_living15    1     10.75  436.99 -68351
## - condition        1     11.28  437.52 -68329
## - sqft_lot         1     11.78  438.02 -68308
## - waterfront       1     20.72  446.96 -67937
## - view             1     28.55  454.79 -67619
## - grade            1     44.60  470.84 -66983
## - sqft_above       1     72.27  498.51 -65936
## - sqft_basement    1    155.49  581.73 -63105
## - zipcode          69    595.95 1022.19 -52904

```

summary(step.p)

```
##
## Call:
## lm(formula = log(price_sqft) ~ date + bedrooms + bathrooms +
##      sqft_lot + floors + waterfront + view + condition + grade +
##      sqft_above + sqft_basement + yr_built + yr_renovated + zipcode +
##      lat + long + sqft_living15, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.52799 -0.09750 -0.00017  0.09589  0.53071
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -3.561e+01  6.904e+00  -5.158 2.53e-07 ***
## date         1.989e-04  1.004e-05  19.808 < 2e-16 ***
## bedrooms    -3.290e-02  1.702e-03 -19.335 < 2e-16 ***
## bathrooms    1.746e-02  2.777e-03   6.287 3.31e-10 ***
## sqft_lot     9.056e-07  4.032e-08  22.459 < 2e-16 ***
## floors      -5.984e-02  3.298e-03 -18.141 < 2e-16 ***
## waterfront1  5.362e-01  1.800e-02  29.788 < 2e-16 ***
## view         6.552e-02  1.874e-03  34.966 < 2e-16 ***
## condition    4.463e-02  2.031e-03  21.975 < 2e-16 ***
## grade        8.447e-02  1.933e-03  43.701 < 2e-16 ***
## sqft_above   -1.865e-04  3.353e-06 -55.629 < 2e-16 ***
## sqft_basement -3.143e-04  3.852e-06 -81.596 < 2e-16 ***
## yr_built     -5.878e-04  6.868e-05  -8.559 < 2e-16 ***
## yr_renovated  3.512e-05  3.131e-06  11.216 < 2e-16 ***
## zipcode98002  7.403e-03  1.477e-02   0.501 0.616177
## zipcode98003  7.173e-03  1.324e-02   0.542 0.587894
## zipcode98004  1.048e+00  2.457e-02  42.668 < 2e-16 ***
## zipcode98005  6.491e-01  2.621e-02  24.766 < 2e-16 ***
## zipcode98006  5.916e-01  2.164e-02  27.344 < 2e-16 ***
## zipcode98007  5.734e-01  2.720e-02  21.084 < 2e-16 ***
## zipcode98008  5.583e-01  2.605e-02  21.432 < 2e-16 ***
## zipcode98010  2.390e-01  2.504e-02   9.544 < 2e-16 ***
## zipcode98011  2.960e-01  3.343e-02   8.855 < 2e-16 ***
## zipcode98014  2.535e-01  3.889e-02   6.518 7.29e-11 ***
## zipcode98019  2.153e-01  3.713e-02   5.798 6.82e-09 ***
## zipcode98022  1.089e-01  2.041e-02   5.336 9.60e-08 ***
## zipcode98023 -4.300e-02  1.233e-02  -3.488 0.000489 ***
## zipcode98024  3.829e-01  3.582e-02  10.688 < 2e-16 ***
## zipcode98027  4.947e-01  2.272e-02  21.778 < 2e-16 ***
## zipcode98028  2.556e-01  3.238e-02   7.895 3.06e-15 ***
## zipcode98029  5.466e-01  2.599e-02  21.030 < 2e-16 ***
## zipcode98030  4.246e-02  1.467e-02   2.895 0.003792 **
## zipcode98031  4.846e-02  1.531e-02   3.166 0.001549 **
## zipcode98032 -2.413e-02  1.830e-02  -1.318 0.187474
## zipcode98033  6.536e-01  2.796e-02  23.377 < 2e-16 ***
## zipcode98034  3.963e-01  2.991e-02  13.252 < 2e-16 ***
## zipcode98038  1.708e-01  1.735e-02   9.842 < 2e-16 ***
## zipcode98039  1.273e+00  3.702e-02  34.385 < 2e-16 ***
## zipcode98040  8.205e-01  2.172e-02  37.771 < 2e-16 ***
```

```

## zipcode98042 6.631e-02 1.449e-02 4.576 4.78e-06 ***
## zipcode98045 3.702e-01 3.319e-02 11.153 < 2e-16 ***
## zipcode98052 5.136e-01 2.879e-02 17.843 < 2e-16 ***
## zipcode98053 4.898e-01 3.125e-02 15.673 < 2e-16 ***
## zipcode98055 1.014e-01 1.722e-02 5.890 3.92e-09 ***
## zipcode98056 2.668e-01 1.874e-02 14.242 < 2e-16 ***
## zipcode98058 1.248e-01 1.640e-02 7.606 2.97e-14 ***
## zipcode98059 2.898e-01 1.852e-02 15.645 < 2e-16 ***
## zipcode98065 3.824e-01 2.995e-02 12.765 < 2e-16 ***
## zipcode98070 2.390e-01 2.357e-02 10.139 < 2e-16 ***
## zipcode98072 3.392e-01 3.350e-02 10.126 < 2e-16 ***
## zipcode98074 4.708e-01 2.761e-02 17.053 < 2e-16 ***
## zipcode98075 4.892e-01 2.663e-02 18.368 < 2e-16 ***
## zipcode98077 3.250e-01 3.505e-02 9.271 < 2e-16 ***
## zipcode98092 2.147e-02 1.328e-02 1.617 0.105960
## zipcode98102 8.733e-01 2.915e-02 29.954 < 2e-16 ***
## zipcode98103 6.964e-01 2.679e-02 25.992 < 2e-16 ***
## zipcode98105 8.189e-01 2.754e-02 29.734 < 2e-16 ***
## zipcode98106 3.190e-01 1.991e-02 16.026 < 2e-16 ***
## zipcode98107 7.191e-01 2.758e-02 26.067 < 2e-16 ***
## zipcode98108 3.088e-01 2.212e-02 13.961 < 2e-16 ***
## zipcode98109 8.527e-01 2.920e-02 29.205 < 2e-16 ***
## zipcode98112 9.307e-01 2.539e-02 36.661 < 2e-16 ***
## zipcode98115 6.882e-01 2.723e-02 25.271 < 2e-16 ***
## zipcode98116 6.715e-01 2.211e-02 30.373 < 2e-16 ***
## zipcode98117 6.809e-01 2.755e-02 24.713 < 2e-16 ***
## zipcode98118 4.086e-01 1.939e-02 21.075 < 2e-16 ***
## zipcode98119 8.518e-01 2.716e-02 31.359 < 2e-16 ***
## zipcode98122 6.993e-01 2.395e-02 29.202 < 2e-16 ***
## zipcode98125 4.234e-01 2.942e-02 14.389 < 2e-16 ***
## zipcode98126 4.892e-01 2.030e-02 24.095 < 2e-16 ***
## zipcode98133 3.188e-01 3.034e-02 10.507 < 2e-16 ***
## zipcode98136 6.093e-01 2.085e-02 29.219 < 2e-16 ***
## zipcode98144 5.846e-01 2.227e-02 26.251 < 2e-16 ***
## zipcode98146 2.177e-01 1.875e-02 11.608 < 2e-16 ***
## zipcode98148 1.484e-01 2.790e-02 5.319 1.05e-07 ***
## zipcode98155 2.845e-01 3.164e-02 8.993 < 2e-16 ***
## zipcode98166 2.649e-01 1.722e-02 15.383 < 2e-16 ***
## zipcode98168 6.312e-02 1.798e-02 3.510 0.000449 ***
## zipcode98177 3.883e-01 3.188e-02 12.179 < 2e-16 ***
## zipcode98178 1.093e-01 1.871e-02 5.840 5.32e-09 ***
## zipcode98188 7.227e-02 1.944e-02 3.717 0.000202 ***
## zipcode98198 3.757e-02 1.464e-02 2.566 0.010302 *
## zipcode98199 7.278e-01 2.619e-02 27.793 < 2e-16 ***
## lat 3.748e-01 6.633e-02 5.650 1.62e-08 ***
## long -1.669e-01 5.277e-02 -3.163 0.001564 **
## sqft_living15 6.636e-05 3.094e-06 21.452 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1528 on 18251 degrees of freedom

```

```
## Multiple R-squared:  0.8348, Adjusted R-squared:  0.8341  
## F-statistic: 1085 on 85 and 18251 DF,  p-value: < 2.2e-16
```

```
pred.step <- exp(predict(step.p, test))  
step.mse <- mean((test$price_sqft - pred.step)^2)  
  
step.mse
```

```
## [1] 2060.988
```

```
train2 <- train[,-1]  
test2 <- test[,-1]  
  
train2$waterfront <- as.numeric(train2$waterfront)  
train2$zipcode <- as.numeric(train2$zipcode)  
  
test2$waterfront <- as.numeric(test2$waterfront)  
test2$zipcode <- as.numeric(test2$zipcode)  
  
opt.lambda <- cv.glmnet(as.matrix(train2[,-18]),  
                        as.matrix(log(train2$price_sqft)),  
                        alpha = 1)$lambda.min  
lasso <- glmnet(as.matrix(train2[,-18]),  
               as.matrix(log(train2$price_sqft)), alpha = 1, lambda = opt.lambda)  
  
coef(lasso)
```

```
## 18 x 1 sparse Matrix of class "dgCMatrix"  
##  
## (Intercept) -7.364280e+01  
## bedrooms -5.062452e-02  
## bathrooms 4.298592e-02  
## sqft_lot 5.561622e-07  
## floors 5.433222e-02  
## waterfront 4.257176e-01  
## view 7.086899e-02  
## condition 4.807474e-02  
## grade 1.544870e-01  
## sqft_above -2.505722e-04  
## sqft_basement -2.800686e-04  
## yr_built -3.849553e-03  
## yr_renovated 2.889652e-05  
## zipcode -1.416472e-03  
## lat 1.403804e+00  
## long -1.521154e-01  
## sqft_living15 7.431047e-05  
## sqft_lot15 -2.119418e-07
```

```
p.lasso <- exp(predict(lasso, s = opt.lambda,
                      newx = as.matrix(test2[,-18])))
lasso.mse <- mean((test2$price_sqft - p.lasso)^2)

lasso.mse
```

```
## [1] 4931.009
```

```
opt.lambda2 <- cv.glmnet(as.matrix(train2[,-18]),
                        as.matrix(log(train2$price_sqft)),
                        alpha = 0)$lambda.min
ridge <- glmnet(as.matrix(train2[,-18]),
               as.matrix(log(train2$price_sqft)), alpha = 0, lambda = opt.lambda2)

coef(ridge)
```

```
## 18 x 1 sparse Matrix of class "dgCMatrix"
##                s0
## (Intercept)  -7.503671e+01
## bedrooms    -6.057232e-02
## bathrooms    1.810302e-02
## sqft_lot      4.389692e-07
## floors       4.839973e-02
## waterfront    4.125402e-01
## view         6.781407e-02
## condition    4.816806e-02
## grade        1.228191e-01
## sqft_above   -1.727866e-04
## sqft_basement -2.142110e-04
## yr_built     -3.226353e-03
## yr_renovated  3.363895e-05
## zipcode     -1.119356e-03
## lat         1.343662e+00
## long        -1.787057e-01
## sqft_living15 5.454289e-05
## sqft_lot15   -2.356043e-07
```

```
p.ridge <- exp(predict(ridge, alpha = 0,
                      newx = as.matrix(test2[,-18])))
ridge.mse <- mean((test2$price_sqft - p.ridge)^2)

ridge.mse
```

```
## [1] 5068.003
```



```
PCR <- pcr(log(price_sqft)~., data = train, scale = TRUE, validation = "CV")  
  
summary(PCR)
```

```

## Data:      X dimension: 18337 86
## Y dimension: 18337 1
## Fit method: svdpc
## Number of components considered: 86
##
## VALIDATION: RMSEP
## Cross-validated using 10 random segments.
##      (Intercept)  1 comps  2 comps  3 comps  4 comps  5 comps  6 comps
## CV              0.3752   0.3716   0.3263   0.326   0.2939   0.286   0.2807
## adjCV           0.3752   0.3716   0.3262   0.326   0.2939   0.286   0.2807
##      7 comps  8 comps  9 comps 10 comps 11 comps 12 comps 13 comps
## CV           0.2627   0.2556   0.2556   0.2545   0.2475   0.2475   0.2449
## adjCV        0.2626   0.2554   0.2555   0.2541   0.2476   0.2477   0.2451
##      14 comps 15 comps 16 comps 17 comps 18 comps 19 comps 20 comps
## CV           0.2446   0.2444   0.2435   0.2426   0.2419   0.2418   0.2416
## adjCV        0.2447   0.2447   0.2437   0.2420   0.2419   0.2420   0.2419
##      21 comps 22 comps 23 comps 24 comps 25 comps 26 comps 27 comps
## CV           0.2411   0.2409   0.2404   0.2403   0.2400   0.2396   0.2395
## adjCV        0.2412   0.2411   0.2405   0.2406   0.2401   0.2397   0.2398
##      28 comps 29 comps 30 comps 31 comps 32 comps 33 comps 34 comps
## CV           0.2379   0.2367   0.2365   0.2362   0.2359   0.2356   0.2354
## adjCV        0.2377   0.2366   0.2367   0.2363   0.2359   0.2354   0.2354
##      35 comps 36 comps 37 comps 38 comps 39 comps 40 comps 41 comps
## CV           0.2353   0.2353   0.2349   0.2344   0.2337   0.2330   0.2323
## adjCV        0.2354   0.2354   0.2347   0.2344   0.2336   0.2329   0.2324
##      42 comps 43 comps 44 comps 45 comps 46 comps 47 comps 48 comps
## CV           0.2319   0.2315   0.2314   0.2311   0.2308   0.2306   0.2305
## adjCV        0.2316   0.2314   0.2316   0.2310   0.2308   0.2306   0.2306
##      49 comps 50 comps 51 comps 52 comps 53 comps 54 comps 55 comps
## CV           0.2301   0.2299   0.2299   0.2295   0.2292   0.2290   0.2287
## adjCV        0.2303   0.2301   0.2300   0.2296   0.2294   0.2291   0.2288
##      56 comps 57 comps 58 comps 59 comps 60 comps 61 comps 62 comps
## CV           0.2287   0.2285   0.2283   0.2280   0.2280   0.2279   0.2276
## adjCV        0.2288   0.2286   0.2284   0.2281   0.2281   0.2281   0.2277
##      63 comps 64 comps 65 comps 66 comps 67 comps 68 comps 69 comps
## CV           0.2276   0.2276   0.2275   0.2274   0.2275   0.2272   0.2262
## adjCV        0.2277   0.2277   0.2276   0.2276   0.2277   0.2273   0.2264
##      70 comps 71 comps 72 comps 73 comps 74 comps 75 comps 76 comps
## CV           0.2014   0.1923   0.1917   0.1904   0.1851   0.1841   0.1825
## adjCV        0.2009   0.1922   0.1916   0.1904   0.1850   0.1841   0.1824
##      77 comps 78 comps 79 comps 80 comps 81 comps 82 comps 83 comps
## CV           0.1824   0.1809   0.1733   0.1732   0.1720   0.1685   0.1582
## adjCV        0.1824   0.1808   0.1732   0.1732   0.1721   0.1685   0.1582
##      84 comps 85 comps 86 comps
## CV           0.1543   0.1543   0.1532
## adjCV        0.1543   0.1543   0.1532
##
## TRAINING: % variance explained
##      1 comps  2 comps  3 comps  4 comps  5 comps  6 comps  7 comps
## X              5.510   8.368   10.85   12.98   14.73   16.26   17.66
## log(price_sqft) 1.908   24.383   24.47   38.69   41.98   44.12   51.05
##      8 comps  9 comps 10 comps 11 comps 12 comps 13 comps

```

## X	18.96	20.25	21.50	22.74	23.95	25.15
## log(price_sqft)	53.67	53.67	54.19	56.62	56.62	57.59
##	14 comps	15 comps	16 comps	17 comps	18 comps	19 comps
## X	26.34	27.54	28.73	29.92	31.12	32.31
## log(price_sqft)	57.71	57.72	58.08	58.64	58.69	58.70
##	20 comps	21 comps	22 comps	23 comps	24 comps	25 comps
## X	33.50	34.69	35.88	37.07	38.25	39.44
## log(price_sqft)	58.73	58.94	59.01	59.21	59.23	59.36
##	26 comps	27 comps	28 comps	29 comps	30 comps	31 comps
## X	40.62	41.81	42.99	44.18	45.36	46.54
## log(price_sqft)	59.52	59.52	60.21	60.54	60.56	60.73
##	32 comps	33 comps	34 comps	35 comps	36 comps	37 comps
## X	47.73	48.91	50.09	51.27	52.45	53.63
## log(price_sqft)	60.84	61.03	61.04	61.04	61.10	61.30
##	38 comps	39 comps	40 comps	41 comps	42 comps	43 comps
## X	54.81	55.99	57.17	58.35	59.52	60.70
## log(price_sqft)	61.37	61.61	61.81	62.00	62.25	62.32
##	44 comps	45 comps	46 comps	47 comps	48 comps	49 comps
## X	61.88	63.06	64.23	65.41	66.59	67.77
## log(price_sqft)	62.32	62.52	62.61	62.70	62.70	62.79
##	50 comps	51 comps	52 comps	53 comps	54 comps	55 comps
## X	68.94	70.12	71.29	72.47	73.64	74.82
## log(price_sqft)	62.85	62.89	62.99	63.06	63.13	63.23
##	56 comps	57 comps	58 comps	59 comps	60 comps	61 comps
## X	75.99	77.16	78.34	79.51	80.68	81.85
## log(price_sqft)	63.23	63.30	63.38	63.48	63.48	63.52
##	62 comps	63 comps	64 comps	65 comps	66 comps	67 comps
## X	83.02	84.19	85.36	86.52	87.69	88.86
## log(price_sqft)	63.62	63.63	63.63	63.66	63.67	63.67
##	68 comps	69 comps	70 comps	71 comps	72 comps	73 comps
## X	90.02	91.13	92.21	93.26	94.29	95.27
## log(price_sqft)	63.77	64.02	71.64	74.02	74.13	74.47
##	74 comps	75 comps	76 comps	77 comps	78 comps	79 comps
## X	96.12	96.94	97.64	98.26	98.75	99.04
## log(price_sqft)	75.89	76.13	76.55	76.56	76.97	78.86
##	80 comps	81 comps	82 comps	83 comps	84 comps	85 comps
## X	99.32	99.57	99.78	99.92	99.98	100.00
## log(price_sqft)	78.88	79.17	80.01	82.37	83.23	83.23
##	86 comps					
## X	100.00					
## log(price_sqft)	83.48					

```
PCR.p <- exp(predict(PCR, test, ncomp = 86))
```

```
PCR.mse <- mean((test$price_sqft - PCR.p)^2)
```

```
PCR.mse
```

```
## [1] 2061.569
```

```

# Random Forest regression CV

# Creating a Mode function
Mode <- function(x) {
  uni <- unique(x)
  uni[which.max(tabulate(match(x, uni)))]
}

# Creating a subsample of the data to CV due to computational strain
set.seed(10)

n2 <- NULL
m2 <- NULL

for (cv in 1:10) {
  rf.sub <- train[sample(nrow(train), nrow(train)/20),]
  cv.flag <- sample(nrow(rf.sub), nrow(rf.sub)/10)
  cv.train <- rf.sub[-cv.flag,]
  cv.test <- rf.sub[cv.flag,]

  n <- NULL
  m <- NULL

  for (i in c(100,200,300,400,500,600,700,800,900,1000)) {

    ms <- NULL

    for (k in 1:10) {
      rf.cv <- randomForest(log(price_sqft)~., data = cv.train[,c(-6,-14)]
                            + as.numeric(cv.train$zipcode)
                            + as.numeric(cv.train$waterfront),
                            ntree = i, mtry = k)
      p.rf <- exp(predict(rf.cv, cv.test))
      cv.mse <- mean((cv.test$price_sqft - p.rf)^2)
      ms <- c(ms, cv.mse)
    }

    min.mse <- min(ms)
    m <- c(m, which.min(ms))
    n <- c(n, min.mse)
  }

  m2 <- c(m2, Mode(m))
  n2 <- c(n2, which.min(n)*100)
}

opt.mtry <- Mode(m2)
opt.ntree <- Mode(n2)

cat("Opt mtry:", opt.mtry, "\n")

```

```
## Opt mtry: 1
```

```
cat("Opt ntrees:", opt.ntree)
```

```
## Opt ntrees: 100
```

```
rf.p <- randomForest(log(price_sqft) ~ ., data = train[,c(-6,-14)]
                    + as.numeric(train$zipcode)
                    + as.numeric(train$waterfront), importance = TRUE,
                    ntree = opt.ntree, mtry = opt.mtry)
```

```
rf.p
```

```
##
## Call:
## randomForest(formula = log(price_sqft) ~ ., data = train[, c(-6,      -14)] + as.numeric(trai
in$zipcode) + as.numeric(train$waterfront),      importance = TRUE, ntree = opt.ntree, mtry = op
t.mtry)
##              Type of random forest: regression
##              Number of trees: 100
## No. of variables tried at each split: 1
##
##              Mean of squared residuals: 0.02431407
##              % Var explained: 80.52
```

```
rf.p$importance
```

```
##              %IncMSE IncNodePurity
## date          0.001437709      46.51502
## bedrooms      0.037876210     174.57475
## bathrooms     0.040465857     170.93269
## sqft_lot      0.010973615      84.11589
## floors        0.049263666     168.57854
## view          0.053970721     154.71353
## condition     0.040382865     143.72134
## grade         0.038540517     171.81303
## sqft_above    0.009847719      75.63696
## sqft_basement 0.025284156     144.15302
## yr_built      0.013835901      77.53221
## yr_renovated  0.042285958     153.57581
## lat           0.053178687     175.07680
## long          0.058847797     199.30080
## sqft_living15 0.010841314      78.75308
## sqft_lot15    0.014116382     109.24761
```

```
pred.rf <- exp(predict(rf.p, test[,c(-6,-14)]
                    + as.numeric(test$zipcode)
                    + as.numeric(test$waterfront)))
rf.mse <- mean((test$price_sqft - pred.rf)^2)

rf.mse
```

```
## [1] 4274.357
```

```
# Generalized additive model
```

```
f <- c(4,10,11,12,13,17,18)
sm <- colnames(train[,f])
nm <- colnames(train[,c(-f, -19)])

v <- paste0("s(", sm, ")", collapse = "+")
n2 <- paste0(nm, collapse = "+")
vn <- paste0(v, "+", n2)

form <- as.formula(paste0("log(price_sqft)~", vn, collapse = ""))
gm <- gam(formula = form, data = train)

summary(gm)
```

```
##
## Family: gaussian
## Link function: identity
##
## Formula:
## log(price_sqft) ~ s(sqft_lot) + s(sqft_above) + s(sqft_basement) +
##   s(yr_built) + s(yr_renovated) + s(sqft_living15) + s(sqft_lot15) +
##   date + bedrooms + bathrooms + floors + waterfront + view +
##   condition + grade + zipcode + lat + long
##
## Parametric coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -4.327e-02 1.079e-02 -4.012 6.05e-05 ***
## date         2.026e-04 9.585e-06 21.132 < 2e-16 ***
## bedrooms    -1.905e-02 1.695e-03 -11.237 < 2e-16 ***
## bathrooms    2.563e-02 2.794e-03  9.175 < 2e-16 ***
## floors      -5.708e-02 3.524e-03 -16.200 < 2e-16 ***
## waterfront1  5.242e-01 1.725e-02 30.379 < 2e-16 ***
## view         6.675e-02 1.814e-03 36.790 < 2e-16 ***
## condition    5.254e-02 1.984e-03 26.489 < 2e-16 ***
## grade        8.781e-02 1.893e-03 46.387 < 2e-16 ***
## zipcode98002 -8.943e-03 1.394e-02 -0.642 0.52105
## zipcode98003  3.861e-02 1.253e-02  3.080 0.00207 **
## zipcode98004  1.058e+00 2.322e-02 45.576 < 2e-16 ***
## zipcode98005  6.792e-01 2.502e-02 27.152 < 2e-16 ***
## zipcode98006  5.878e-01 2.076e-02 28.308 < 2e-16 ***
## zipcode98007  5.997e-01 2.606e-02 23.013 < 2e-16 ***
## zipcode98008  5.830e-01 2.499e-02 23.329 < 2e-16 ***
## zipcode98010  1.627e-01 2.110e-02  7.711 1.31e-14 ***
## zipcode98011  3.426e-01 3.123e-02 10.971 < 2e-16 ***
## zipcode98014  1.717e-01 3.601e-02  4.769 1.87e-06 ***
## zipcode98019  1.796e-01 3.533e-02  5.084 3.74e-07 ***
## zipcode98022  2.305e-02 1.327e-02  1.737 0.08240 .
## zipcode98023  1.325e-04 1.109e-02  0.012 0.99047
## zipcode98024  2.838e-01 3.279e-02  8.655 < 2e-16 ***
## zipcode98027  4.691e-01 2.114e-02 22.190 < 2e-16 ***
## zipcode98028  3.073e-01 2.963e-02 10.370 < 2e-16 ***
## zipcode98029  5.229e-01 2.419e-02 21.620 < 2e-16 ***
## zipcode98030  3.484e-02 1.386e-02  2.513 0.01197 *
## zipcode98031  5.513e-02 1.460e-02  3.776 0.00016 ***
## zipcode98032 -1.754e-03 1.742e-02 -0.101 0.91982
## zipcode98033  6.790e-01 2.637e-02 25.749 < 2e-16 ***
## zipcode98034  4.527e-01 2.789e-02 16.230 < 2e-16 ***
## zipcode98038  1.221e-01 1.381e-02  8.840 < 2e-16 ***
## zipcode98039  1.284e+00 3.501e-02 36.689 < 2e-16 ***
## zipcode98040  8.457e-01 2.052e-02 41.209 < 2e-16 ***
## zipcode98042  3.642e-02 1.268e-02  2.872 0.00408 **
## zipcode98045  2.472e-01 2.543e-02  9.719 < 2e-16 ***
## zipcode98052  5.344e-01 2.755e-02 19.397 < 2e-16 ***
## zipcode98053  4.659e-01 2.982e-02 15.622 < 2e-16 ***
## zipcode98055  1.116e-01 1.649e-02  6.767 1.35e-11 ***
## zipcode98056  2.654e-01 1.797e-02 14.774 < 2e-16 ***
```

```

## zipcode98058 1.277e-01 1.554e-02 8.218 < 2e-16 ***
## zipcode98059 2.750e-01 1.764e-02 15.593 < 2e-16 ***
## zipcode98065 2.981e-01 2.549e-02 11.697 < 2e-16 ***
## zipcode98070 2.917e-01 1.964e-02 14.851 < 2e-16 ***
## zipcode98072 3.679e-01 3.185e-02 11.551 < 2e-16 ***
## zipcode98074 4.500e-01 2.633e-02 17.091 < 2e-16 ***
## zipcode98075 4.524e-01 2.508e-02 18.034 < 2e-16 ***
## zipcode98077 3.252e-01 3.367e-02 9.660 < 2e-16 ***
## zipcode98092 3.883e-03 1.187e-02 0.327 0.74355
## zipcode98102 9.246e-01 2.640e-02 35.018 < 2e-16 ***
## zipcode98103 7.518e-01 2.305e-02 32.613 < 2e-16 ***
## zipcode98105 8.639e-01 2.494e-02 34.638 < 2e-16 ***
## zipcode98106 3.338e-01 1.697e-02 19.674 < 2e-16 ***
## zipcode98107 7.754e-01 2.324e-02 33.363 < 2e-16 ***
## zipcode98108 3.336e-01 2.010e-02 16.596 < 2e-16 ***
## zipcode98109 9.126e-01 2.595e-02 35.169 < 2e-16 ***
## zipcode98112 9.823e-01 2.299e-02 42.723 < 2e-16 ***
## zipcode98115 7.350e-01 2.426e-02 30.293 < 2e-16 ***
## zipcode98116 7.239e-01 1.799e-02 40.248 < 2e-16 ***
## zipcode98117 7.380e-01 2.289e-02 32.240 < 2e-16 ***
## zipcode98118 4.239e-01 1.787e-02 23.722 < 2e-16 ***
## zipcode98119 9.172e-01 2.343e-02 39.153 < 2e-16 ***
## zipcode98122 7.437e-01 2.174e-02 34.205 < 2e-16 ***
## zipcode98125 4.790e-01 2.613e-02 18.333 < 2e-16 ***
## zipcode98126 5.146e-01 1.690e-02 30.445 < 2e-16 ***
## zipcode98133 3.833e-01 2.601e-02 14.736 < 2e-16 ***
## zipcode98136 6.561e-01 1.717e-02 38.213 < 2e-16 ***
## zipcode98144 6.112e-01 2.019e-02 30.271 < 2e-16 ***
## zipcode98146 2.470e-01 1.598e-02 15.456 < 2e-16 ***
## zipcode98148 1.896e-01 2.613e-02 7.254 4.20e-13 ***
## zipcode98155 3.373e-01 2.798e-02 12.052 < 2e-16 ***
## zipcode98166 2.979e-01 1.504e-02 19.815 < 2e-16 ***
## zipcode98168 6.890e-02 1.626e-02 4.238 2.27e-05 ***
## zipcode98177 4.810e-01 2.700e-02 17.819 < 2e-16 ***
## zipcode98178 1.139e-01 1.772e-02 6.428 1.32e-10 ***
## zipcode98188 9.144e-02 1.826e-02 5.006 5.60e-07 ***
## zipcode98198 7.612e-02 1.349e-02 5.642 1.71e-08 ***
## zipcode98199 8.054e-01 2.145e-02 37.539 < 2e-16 ***
## lat 2.692e-01 5.915e-02 4.551 5.36e-06 ***
## long 9.632e-02 2.290e-02 4.206 2.61e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
##           edf Ref.df      F p-value
## s(sqft_lot)      8.704  8.968  35.934 <2e-16 ***
## s(sqft_above)     6.415  7.467 612.898 <2e-16 ***
## s(sqft_basement)  4.021  4.948 1520.126 <2e-16 ***
## s(yr_built)       8.326  8.870  69.876 <2e-16 ***
## s(yr_renovated)   3.162  3.473  76.895 <2e-16 ***
## s(sqft_living15)  5.055  6.196  65.277 <2e-16 ***
## s(sqft_lot15)     8.747  8.980   6.598 <2e-16 ***

```



```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Rank: 138/143
## R-sq.(adj) =  0.849   Deviance explained =   85%
## GCV = 0.021388   Scale est. = 0.021244   n = 18337
```

```
pred.gm <- exp(predict(gm, test))
gm.mse <- mean((test$price_sqft - pred.gm)^2)

gm.mse
```

```
## [1] 1887.063
```

```
# Averaging predictions of all models together
combined <- (pred.lm6 + pred.step + pred.rf + pred.gm + p.lasso[,1] +
             p.ridge[,1] + as.vector(PCR.p)) / 7

combined.mse <- mean((test$price_sqft - combined)^2)

combined.mse
```

```
## [1] 2245.721
```

```
mses <- c(lm6.mse, step.mse, lasso.mse, ridge.mse, PCR.mse, rf.mse, gm.mse,
          combined.mse)

MSEs <- data.frame(Model = c("Linear Model", "Stepwise", "LASSO", "Ridge",
                             "PCR", "Random Forest", "GAM", "Combination"),
                  MSE = mses)

MSEs
```

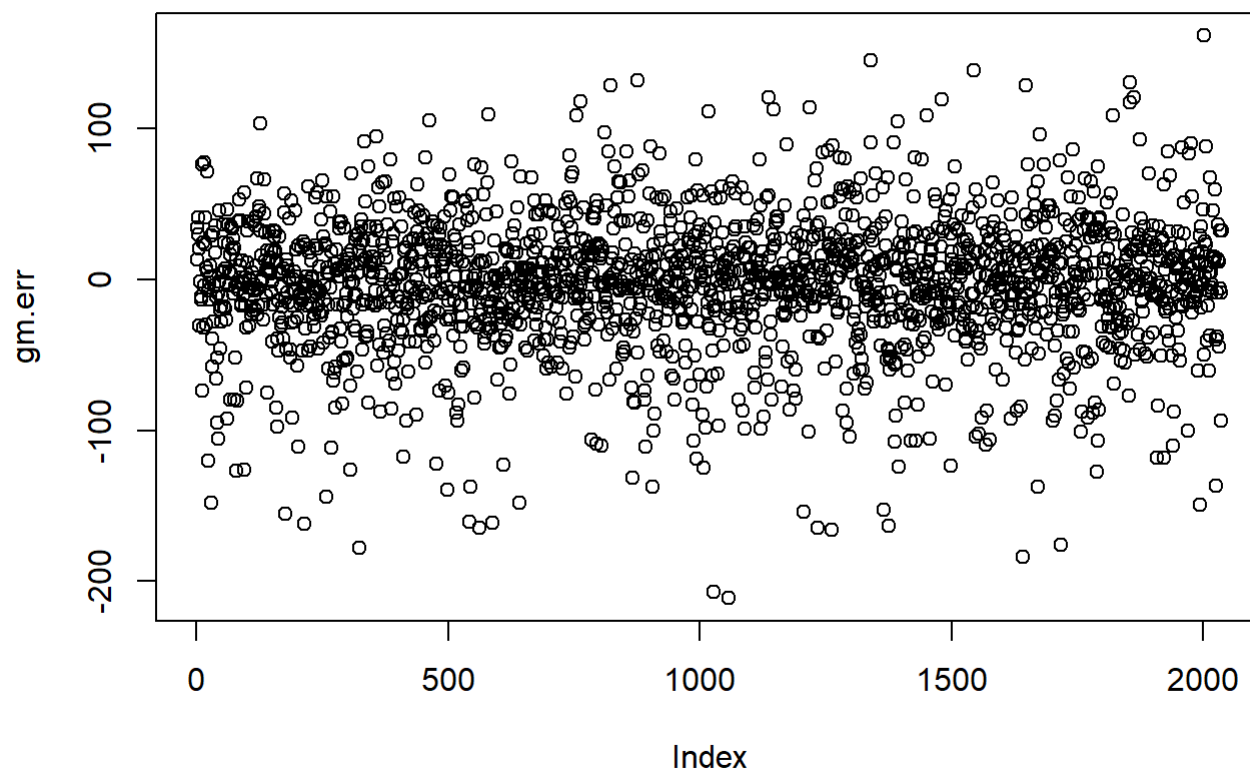
```
##           Model      MSE
## 1 Linear Model 2061.513
## 2      Stepwise 2060.988
## 3         LASSO 4931.009
## 4         Ridge 5068.003
## 5          PCR 2061.569
## 6 Random Forest 4274.357
## 7          GAM 1887.063
## 8 Combination 2245.721
```

```
gm.err <- pred.gm - test$price_sqft

summary(gm.err)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.    Max.
## -211.3528 -22.2442   -0.6779   -2.9227   21.0632   161.7432
```

```
plot(gm.err)
```



```
test[which.max(abs(gm.err)),]
```

```
##      date bedrooms bathrooms sqft_lot floors waterfront view condition
## 10695 2015-02-25      3      1   6120   1.5          0  0          3
##      grade sqft_above sqft_basement yr_built yr_renovated zipcode    lat
## 10695    7    1140          0    1926          0  98115 47.6822
##      long sqft_living15 sqft_lot15 price_sqft
## 10695 -122.309      1800      4080   617.5439
```

```
test[which.max(gm.err),]
```

```
##          date bedrooms bathrooms sqft_lot floors waterfront view condition
## 3988 2014-07-01          1          1      833      1          0      0          4
##      grade sqft_above sqft_basement yr_built yr_renovated zipcode      lat
## 3988      7          590              0      1926              0  98122 47.6082
##      long sqft_living15 sqft_lot15 price_sqft
## 3988 -122.299          780          1617   342.3729
```