

Assignment: ASSIGNMENT 8, Housing data analysis

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```
# load readxl library and set directory
# Read the housing data set
library(readxl)
setwd("/Users/mshekhar/Desktop/R Programming/DSC520/stats_for_data_science/stats_for_data_science")
mydata <- read_excel("week-6-housing.xlsx")

# check the structure of data and some basic stats
str(mydata)

## tibble[,24] [12,865 x 24] (S3: tbl_df/tbl/data.frame)
##  $ Sale Date           : POSIXct[1:12865], format: "2006-01-03" "2006-01-03" ...
##  $ Sale Price          : num [1:12865] 698000 649990 572500 420000 369900 ...
##  $ sale_reason         : num [1:12865] 1 1 1 1 1 1 1 1 1 1 ...
##  $ sale_instrument     : num [1:12865] 3 3 3 3 3 15 3 3 3 3 ...
##  $ sale_warning        : chr [1:12865] NA NA NA NA ...
##  $ sitetype            : chr [1:12865] "R1" "R1" "R1" "R1" ...
##  $ addr_full           : chr [1:12865] "17021 NE 113TH CT" "11927 178TH PL NE" "13315 174TH AVE I
##  $ zip5                : num [1:12865] 98052 98052 98052 98052 98052 ...
##  $ ctynome             : chr [1:12865] "REDMOND" "REDMOND" NA "REDMOND" ...
##  $ postalctyn         : chr [1:12865] "REDMOND" "REDMOND" "REDMOND" "REDMOND" ...
##  $ lon                 : num [1:12865] -122 -122 -122 -122 -122 ...
##  $ lat                 : num [1:12865] 47.7 47.7 47.7 47.6 47.7 ...
##  $ building_grade      : num [1:12865] 9 9 8 8 7 7 10 10 9 8 ...
##  $ square_feet_total_living: num [1:12865] 2810 2880 2770 1620 1440 4160 3960 3720 4160 2760 ...
##  $ bedrooms            : num [1:12865] 4 4 4 3 3 4 5 4 4 4 ...
##  $ bath_full_count     : num [1:12865] 2 2 1 1 1 2 3 2 2 1 ...
##  $ bath_half_count     : num [1:12865] 1 0 1 0 0 1 0 1 1 0 ...
##  $ bath_3qtr_count     : num [1:12865] 0 1 1 1 1 1 1 0 1 1 ...
##  $ year_built          : num [1:12865] 2003 2006 1987 1968 1980 ...
##  $ year_renovated      : num [1:12865] 0 0 0 0 0 0 0 0 0 0 ...
##  $ current_zoning      : chr [1:12865] "R4" "R4" "R6" "R4" ...
##  $ sq_ft_lot           : num [1:12865] 6635 5570 8444 9600 7526 ...
##  $ prop_type           : chr [1:12865] "R" "R" "R" "R" ...
##  $ present_use         : num [1:12865] 2 2 2 2 2 2 2 2 2 2 ...

summary(mydata)

##      Sale Date           Sale Price      sale_reason
##  Min.   :2006-01-03 00:00:00   Min.    :   698   Min.    : 0.00
##  1st Qu.:2008-07-07 00:00:00   1st Qu.: 460000   1st Qu.: 1.00
##  Median :2011-11-17 00:00:00   Median : 593000   Median : 1.00
##  Mean   :2011-07-28 15:07:32   Mean    : 660738   Mean    : 1.55
##  3rd Qu.:2014-06-05 00:00:00   3rd Qu.: 750000   3rd Qu.: 1.00
##  Max.   :2016-12-16 00:00:00   Max.    :4400000   Max.    :19.00
##  sale_instrument  sale_warning      sitetype      addr_full
##  Min.    : 0.000   Length:12865   Length:12865   Length:12865
##  1st Qu.: 3.000   Class :character   Class :character   Class :character
##  Median : 3.000   Mode  :character   Mode  :character   Mode  :character
```

```
## Mean      : 3.678
## 3rd Qu.: 3.000
## Max.      :27.000
##      zip5      ctyname      postalctyn      lon
## Min.      :98052 Length:12865 Length:12865 Min.      : -122.2
## 1st Qu.:98052 Class :character Class :character 1st Qu.: -122.1
## Median :98052 Mode  :character Mode  :character Median : -122.1
## Mean      :98053 Mean      : -122.1
## 3rd Qu.:98053 3rd Qu.: -122.0
## Max.      :98074 Max.      : -121.9
##      lat      building_grade square_feet_total_living bedrooms
## Min.      :47.46 Min.      : 2.00 Min.      : 240 Min.      : 0.000
## 1st Qu.:47.67 1st Qu.: 8.00 1st Qu.: 1820 1st Qu.: 3.000
## Median :47.69 Median : 8.00 Median : 2420 Median : 4.000
## Mean      :47.68 Mean      : 8.24 Mean      : 2540 Mean      : 3.479
## 3rd Qu.:47.70 3rd Qu.: 9.00 3rd Qu.: 3110 3rd Qu.: 4.000
## Max.      :47.73 Max.      :13.00 Max.      :13540 Max.      :11.000
## bath_full_count bath_half_count bath_3qtr_count year_built
## Min.      : 0.000 Min.      :0.0000 Min.      :0.000 Min.      :1900
## 1st Qu.: 1.000 1st Qu.:0.0000 1st Qu.:0.000 1st Qu.:1979
## Median : 2.000 Median :1.0000 Median :0.000 Median :1998
## Mean      : 1.798 Mean      :0.6134 Mean      :0.494 Mean      :1993
## 3rd Qu.: 2.000 3rd Qu.:1.0000 3rd Qu.:1.000 3rd Qu.:2007
## Max.      :23.000 Max.      :8.0000 Max.      :8.000 Max.      :2016
## year_renovated current_zoning sq_ft_lot prop_type
## Min.      : 0.00 Length:12865 Min.      : 785 Length:12865
## 1st Qu.: 0.00 Class :character 1st Qu.: 5355 Class :character
## Median : 0.00 Mode  :character Median : 7965 Mode  :character
## Mean      : 26.24 Mean      : 22229
## 3rd Qu.: 0.00 3rd Qu.: 12632
## Max.      :2016.00 Max.      :1631322
## present_use
## Min.      : 0.000
## 1st Qu.: 2.000
## Median : 2.000
## Mean      : 6.598
## 3rd Qu.: 2.000
## Max.      :300.000
```

b.i. Explain any transformations or modifications you made to the dataset

```
## ----- 1. Changed the column name to remove spaces -----
# Change the column names to remove spaces
colnames(mydata)[1] <- "Sale_Date"
colnames(mydata)[2] <- "Sale_Price"

## ----- 2. Checking for NAs in the data -----
apply(mydata, 2, function(x) any(is.na(x) | is.infinite(x)))
```

```
##      Sale_Date      Sale_Price      sale_reason
##      FALSE      FALSE      FALSE
##      sale_instrument      sale_warning      sitetype
```

```

##             FALSE             TRUE             FALSE
##             addr_full          zip5             ctyname
##             FALSE             FALSE             TRUE
##             postalctyn          lon              lat
##             FALSE             FALSE             FALSE
##             building_grade square_feet_total_living bedrooms
##             FALSE             FALSE             FALSE
##             bath_full_count    bath_half_count    bath_3qtr_count
##             FALSE             FALSE             FALSE
##             year_built         year_renovated     current_zoning
##             FALSE             FALSE             FALSE
##             sq_ft_lot          prop_type          present_use
##             FALSE             FALSE             FALSE

# check if there are any NAs in sale_warning
sum(is.na(mydata$sale_warning))

## [1] 10568

# check if there are any NAs in ctyname
sum(is.na(mydata$ctyname))

## [1] 6078

# We found that sale_warning = 10568 and ctyname = 6078 have NAs
# sale_warning sounds like a bad remark and if it is not present.
# Deriving sale_warning indicator and adding it to the data
mydata <- cbind(mydata, "Sale_Warning_in" = as.numeric(!(is.na(mydata$sale_warning))))

# ctyname
# Not worried about this field as we have other attributes like zip code,
# latitude, longitude to predict sale price.
# We can ignore this field while model creation

## ----- 3. changing categorical variables to factors -----
# changing sitetype to factor
# check distinct values
unique(mydata$sitetype)

## [1] "R1" "R2" "R3" "DV" "A1" "R4" "C1"

mydata$sitetype <- factor(mydata$sitetype, labels = c(1,2,3,4,5,6,7))

# changing current zoning to factor
# check current zoning unique values and convert to factor
unique(mydata$current_zoning)

## [1] "R4"      "R6"      "URPSO"   "RA5"     "R3"      "R5"      "RA2.5"
## [8] "RA10"    "R12"     "RA5P"    "R1"      "RA2.5SO" "RA2.5P"  "R4/C"
## [15] "EH"      "R1P"     "BC"      "R8"      "A10"     "R6/C"    "R18"
## [22] "A10SO"   "RA10P"   "GC"

mydata$current_zoning <- factor(mydata$current_zoning, labels = c(1:24))

# check unique values in prop_type
unique(mydata$prop_type)

## [1] "R"

```

```

# As all the values are R, really this variable won't be a good predictor.
# Ignoring it

# don't need addr_full in the final data set as we have zip, latitude, longitude
# ignoring cityname and postalctyyn for the same reason

library(dplyr)

##
## Attaching package: 'dplyr'

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

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

# picking only numerical attributes to check correlation to sale_price
hs_data_prccd_num <- mydata %>% dplyr::select("Sale_Price", "sale_reason", "Sale_Warning_in",
"sale_instrument", "zip5", "lon", "lat", "building_grade",
"square_feet_total_living", "bedrooms", "bath_full_count",
"bath_half_count", "bath_3qtr_count", "year_built",
"year_renovated", "sq_ft_lot", "present_use")

# create correlation matrix
cor(hs_data_prccd_num, method = "pearson", use = "complete.obs")

##           Sale_Price  sale_reason Sale_Warning_in
## Sale_Price          1.00000000 -0.116643537      0.083211139
## sale_reason        -0.11664354  1.000000000      0.355583265
## Sale_Warning_in      0.08321114  0.355583265      1.000000000
## sale_instrument     -0.04070601  0.399856649      0.343648065
## zip5                 0.06014866  0.001393069      0.029425761
## lon                  0.04684745 -0.013531540      0.036575404
## lat                  0.02119829 -0.031965818     -0.034812917
## building_grade       0.39122909 -0.078912289     -0.063020490
## square_feet_total_living 0.45458758 -0.065212764     -0.012080335
## bedrooms             0.22546748 -0.056284214     -0.007526779
## bath_full_count       0.28484899 -0.073127789     -0.015603326
## bath_half_count       0.16582843 -0.051526350     -0.016935167
## bath_3qtr_count       0.03574175 -0.000687221     -0.021751887
## year_built            0.24267127 -0.120238360     -0.052099288
## year_renovated        0.03286429  0.024343886      0.040703002
## sq_ft_lot             0.11981223  0.034005560      0.086937090
## present_use          -0.02542926 -0.002062742      0.003681498
##           sale_instrument      zip5      lon      lat
## Sale_Price        -0.040706012  0.060148664  0.04684745  0.02119829
## sale_reason         0.399856649  0.001393069 -0.01353154 -0.03196582
## Sale_Warning_in     0.343648065  0.029425761  0.03657540 -0.03481292
## sale_instrument      1.000000000 -0.016103809 -0.02192079 -0.04966602
## zip5                -0.016103809  1.000000000  0.35337066 -0.13431164
## lon                 -0.021920789  0.353370662  1.00000000 -0.04101493
## lat                 -0.049666024 -0.134311637 -0.04101493  1.00000000
## building_grade      -0.048442696  0.103989081  0.03312593  0.01217953

```

## square_feet_total_living	-0.007000713	0.095658773	0.06857118	-0.06383585
## bedrooms	-0.001834812	-0.036622671	-0.24149623	-0.15481695
## bath_full_count	-0.025098358	0.098244669	0.13139051	0.07757143
## bath_half_count	-0.034397435	0.024996235	0.02574174	-0.04252565
## bath_3qtr_count	0.015641257	-0.074930308	-0.13644355	-0.10044764
## year_built	-0.096159040	0.100989530	0.34235695	0.37787153
## year_renovated	0.011207272	-0.005762662	-0.02017440	-0.09174116
## sq_ft_lot	0.124543302	0.089844639	0.20108459	-0.14545622
## present_use	0.001793940	0.029112870	0.12844704	0.01760738
##	building_grade	square_feet_total_living	bedrooms	
## Sale_Price	0.39122909		0.454587585	0.225467478
## sale_reason	-0.07891229		-0.065212764	-0.056284214
## Sale_Warning_in	-0.06302049		-0.012080335	-0.007526779
## sale_instrument	-0.04844270		-0.007000713	-0.001834812
## zip5	0.10398908		0.095658773	-0.036622671
## lon	0.03312593		0.068571180	-0.241496234
## lat	0.01217953		-0.063835847	-0.154816948
## building_grade	1.00000000		0.745180518	0.342565614
## square_feet_total_living	0.74518052		1.000000000	0.575347048
## bedrooms	0.34256561		0.575347048	1.000000000
## bath_full_count	0.45941631		0.517783777	0.304598944
## bath_half_count	0.28076582		0.316856642	0.162334411
## bath_3qtr_count	0.06951439		0.198673964	0.247340383
## year_built	0.36192968		0.306427289	0.012441913
## year_renovated	-0.01902482		0.043396050	0.014923912
## sq_ft_lot	0.13843064		0.234104503	0.047173368
## present_use	0.04472257		0.032493682	-0.038895101
##	bath_full_count	bath_half_count	bath_3qtr_count	
## Sale_Price	0.28484899	0.16582843	0.035741748	
## sale_reason	-0.07312779	-0.05152635	-0.000687221	
## Sale_Warning_in	-0.01560333	-0.01693517	-0.021751887	
## sale_instrument	-0.02509836	-0.03439744	0.015641257	
## zip5	0.09824467	0.02499624	-0.074930308	
## lon	0.13139051	0.02574174	-0.136443549	
## lat	0.07757143	-0.04252565	-0.100447636	
## building_grade	0.45941631	0.28076582	0.069514388	
## square_feet_total_living	0.51778378	0.31685664	0.198673964	
## bedrooms	0.30459894	0.16233441	0.247340383	
## bath_full_count	1.00000000	0.21731460	-0.379105667	
## bath_half_count	0.21731460	1.00000000	-0.315842525	
## bath_3qtr_count	-0.37910567	-0.31584253	1.000000000	
## year_built	0.45259966	0.19849009	-0.152354857	
## year_renovated	0.02218095	-0.02665661	0.018358504	
## sq_ft_lot	0.04699255	0.02810463	0.049156430	
## present_use	0.03364950	0.01931299	0.001521253	
##	year_built	year_renovated	sq_ft_lot	present_use
## Sale_Price	0.24267127	0.032864291	0.11981223	-0.025429262
## sale_reason	-0.12023836	0.024343886	0.03400556	-0.002062742
## Sale_Warning_in	-0.05209929	0.040703002	0.08693709	0.003681498
## sale_instrument	-0.09615904	0.011207272	0.12454330	0.001793940
## zip5	0.10098953	-0.005762662	0.08984464	0.029112870
## lon	0.34235695	-0.020174395	0.20108459	0.128447040
## lat	0.37787153	-0.091741159	-0.14545622	0.017607381
## building_grade	0.36192968	-0.019024817	0.13843064	0.044722573

```
## square_feet_total_living 0.30642729 0.043396050 0.23410450 0.032493682
## bedrooms 0.01244191 0.014923912 0.04717337 -0.038895101
## bath_full_count 0.45259966 0.022180945 0.04699255 0.033649501
## bath_half_count 0.19849009 -0.026656615 0.02810463 0.019312989
## bath_3qtr_count -0.15235486 0.018358504 0.04915643 0.001521253
## year_built 1.00000000 -0.224586183 -0.13491395 0.130783379
## year_renovated -0.22458618 1.000000000 0.06320824 -0.017558495
## sq_ft_lot -0.13491395 0.063208244 1.00000000 0.059320852
## present_use 0.13078338 -0.017558495 0.05932085 1.000000000
```

```
# We can see some of the lowest correlations to sale_price are -
# 1. sale_price ~ sale_instrument = -0.040706012
# 2. sale_price ~ zip5 = 0.060148664
# 3. sale_price ~ log = 0.04684745
# 4. sale_price ~ lat = 0.02119829
# 5. sale_price ~ bath_3qtr_count = 0.035741748
# 6. sale_price ~ year_renovated = 0.032864291
# 7. sale_price ~ present_use = -0.025429262
# 8. sale_price ~ sale_warning_in = 0.083211139
```

b. ii. Create two variables; one that will contain the variables Sale Price

and Square Foot of Lot (same variables used from previous assignment on simple

regression) and one that will contain Sale Price and several additional

predictors of your choice. Explain the basis for your additional predictor

selections.

```
# creating model with just sale price and square feet of the sq_ft_lot
hs_data_smpl_rgr_mdl <- lm(Sale_Price ~ sq_ft_lot, data = mydata)

# creating data set with several additional predictors of my choice
# My choice is purely based on their correlation
# with predicted variable Sale_Price
# Highest correlations are -
# 1. Sale_Price ~ sale_reason = -0.116643537
# 2. Sale_Price ~ building_grade = 0.39122909
# 3. Sale_Price ~ square_feet_total_living = 0.454587585
# 4. Sale_Price ~ bedrooms = 0.225467478
# 5. Sale_Price ~ bath_full_count = 0.28484899
# 6. Sale_Price ~ bath_half_count = 0.16582843
# 7. Sale_Price ~ year_built = 0.24267127
# 8. Sale_Price ~ sq_ft_lot = 0.11981223
```

```
# Checking correlation of Sale_Price with factor variables - sitetype, current_zoning
# applying Kruskal-Wallis chi-squared test as categorical variable is not
# dichotomous
```

```
kruskal.test(mydata$Sale_Price~mydata$sitetype)
```

```
##
```

```
## Kruskal-Wallis rank sum test
```

```
##
```

```
## data: mydata$Sale_Price by mydata$sitetype
```

```
## Kruskal-Wallis chi-squared = 387.31, df = 6, p-value < 2.2e-16
```

```
kruskal.test(mydata$Sale_Price~mydata$current_zoning)
```

```
##
```

```
## Kruskal-Wallis rank sum test
```

```
##
```

```
## data: mydata$Sale_Price by mydata$current_zoning
```

```
## Kruskal-Wallis chi-squared = 1636.5, df = 23, p-value < 2.2e-16
```

```
# Both the variables are highly correlated to Sale_Price and will include
# them as well as the predictor variables
```

```
# Run 1 : creating multiple regression model with sale price and several
# other predictors found relevant based on correlation exercise
```

```
hs_data_multi_rgr_mdl <- lm(Sale_Price ~ sale_reason+building_grade+
square_feet_total_living+bedrooms+bath_full_count+bath_half_count+
year_built+sq_ft_lot+sitetype+current_zoning, data = mydata)
```

b. iii. Execute a summary() function on two variables defined in the previous

step to compare the model results. What are the R2 and Adjusted R2 statistics?

Explain what these results tell you about the overall model. Did the inclusion

of the additional predictors help explain any large variations found in Sale Price?

```
# check summary of simple regression model
summary(hs_data_smpl_rgr_mdl)
```

```
##
```

```
## Call:
```

```
## lm(formula = Sale_Price ~ sq_ft_lot, data = mydata)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
## -2016064 -194842  -63293   91565  3735109
```

```
##
```

```

## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) 6.418e+05  3.800e+03  168.90  <2e-16 ***
## sq_ft_lot   8.510e-01  6.217e-02   13.69  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 401500 on 12863 degrees of freedom
## Multiple R-squared:  0.01435,    Adjusted R-squared:  0.01428
## F-statistic: 187.3 on 1 and 12863 DF,  p-value: < 2.2e-16
#Multiple R-squared:  0.01435,    Adjusted R-squared:  0.01428
#F-statistic: 187.3 on 1 and 12863 DF,  p-value: < 2.2e-16

# check summary of multiple regression model
summary(hs_data_multi_rgr_md1)

##
## Call:
## lm(formula = Sale_Price ~ sale_reason + building_grade + square_feet_total_living +
##       bedrooms + bath_full_count + bath_half_count + year_built +
##       sq_ft_lot + sitetype + current_zoning, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2045030  -121627   -41643    47525   3657041
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -1.848e+06  5.584e+05  -3.310  0.000935 ***
## sale_reason     -1.133e+04  1.165e+03  -9.724  < 2e-16 ***
## building_grade    2.596e+04  4.704e+03   5.520  3.46e-08 ***
## square_feet_total_living 1.456e+02  6.064e+00  24.003  < 2e-16 ***
## bedrooms        -1.906e+04  4.867e+03  -3.917  9.03e-05 ***
## bath_full_count    1.533e+04  6.027e+03   2.544  0.010973 *
## bath_half_count    1.924e+04  6.348e+03   3.031  0.002445 **
## year_built        1.429e+03  2.725e+02   5.243  1.61e-07 ***
## sq_ft_lot         2.695e-01  7.200e-02   3.743  0.000182 ***
## sitetype2         8.307e+04  2.380e+05   0.349  0.727110
## sitetype3        -3.429e+05  1.901e+05  -1.804  0.071244 .
## sitetype4        -2.182e+05  1.244e+05  -1.754  0.079421 .
## sitetype5        -3.186e+05  1.257e+05  -2.536  0.011234 *
## sitetype6        -3.619e+05  1.558e+05  -2.323  0.020185 *
## sitetype7        -3.799e+05  3.712e+05  -1.023  0.306113
## current_zoning2    2.891e+06  3.806e+05   7.596  3.26e-14 ***
## current_zoning3   -3.569e+05  3.746e+05  -0.953  0.340767
## current_zoning4   -7.066e+05  1.436e+05  -4.920  8.77e-07 ***
## current_zoning5   -7.278e+05  3.747e+05  -1.943  0.052094 .
## current_zoning6   -6.597e+05  1.368e+05  -4.824  1.42e-06 ***
## current_zoning7   -7.086e+05  1.363e+05  -5.198  2.04e-07 ***
## current_zoning8   -8.296e+05  1.568e+05  -5.290  1.25e-07 ***
## current_zoning9   -7.010e+05  1.881e+05  -3.727  0.000194 ***
## current_zoning10  -6.526e+05  1.354e+05  -4.821  1.44e-06 ***
## current_zoning11  -6.094e+05  1.343e+05  -4.537  5.76e-06 ***
## current_zoning12  -6.255e+05  1.399e+05  -4.472  7.83e-06 ***

```



```

## current_zoning13      -6.927e+05  1.343e+05  -5.158 2.54e-07 ***
## current_zoning14      -7.019e+05  1.343e+05  -5.225 1.77e-07 ***
## current_zoning15      -1.705e+05  1.509e+05  -1.130 0.258483
## current_zoning16      -2.342e+05  1.430e+05  -1.638 0.101544
## current_zoning17      -7.102e+05  1.406e+05  -5.049 4.49e-07 ***
## current_zoning18      -2.589e+05  2.454e+05  -1.055 0.291434
## current_zoning19      -6.443e+05  1.354e+05  -4.758 1.98e-06 ***
## current_zoning20      -7.247e+05  1.669e+05  -4.343 1.42e-05 ***
## current_zoning21      -6.687e+05  1.462e+05  -4.574 4.83e-06 ***
## current_zoning22      -7.133e+05  1.337e+05  -5.336 9.66e-08 ***
## current_zoning23      -7.711e+05  1.399e+05  -5.512 3.61e-08 ***
## current_zoning24      -6.504e+05  1.346e+05  -4.832 1.37e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 349500 on 12827 degrees of freedom
## Multiple R-squared:  0.255, Adjusted R-squared:  0.2528
## F-statistic: 118.7 on 37 and 12827 DF,  p-value: < 2.2e-16

# Run 2 : Recreating multiple regression model by applying backward elimination
# of non significant variables. sitetype appears non-significant
# with p-values > 0.05
hs_data_multi_rgr_mdl_2 <- lm(Sale_Price ~ sale_reason+building_grade+
square_feet_total_living+bedrooms+bath_full_count+bath_half_count+
year_built+sq_ft_lot+current_zoning, data = mydata)
# check summary of multiple regression model
summary(hs_data_multi_rgr_mdl_2)

##
## Call:
## lm(formula = Sale_Price ~ sale_reason + building_grade + square_feet_total_living +
##     bedrooms + bath_full_count + bath_half_count + year_built +
##     sq_ft_lot + current_zoning, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2058754  -122967   -42250    47356   3661440
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.991e+06  5.425e+05  -3.671 0.000243 ***
## sale_reason    -1.147e+04  1.166e+03  -9.835 < 2e-16 ***
## building_grade  2.609e+04  4.699e+03   5.552 2.87e-08 ***
## square_feet_total_living 1.474e+02  6.057e+00  24.328 < 2e-16 ***
## bedrooms     -1.721e+04  4.853e+03  -3.545 0.000393 ***
## bath_full_count  1.441e+04  6.027e+03   2.390 0.016863 *
## bath_half_count  1.736e+04  6.341e+03   2.737 0.006205 **
## year_built     1.385e+03  2.721e+02   5.089 3.66e-07 ***
## sq_ft_lot       2.721e-01  7.176e-02   3.792 0.000150 ***
## current_zoning2  2.888e+06  3.810e+05   7.580 3.68e-14 ***
## current_zoning3  -3.542e+05  3.751e+05  -0.945 0.344921
## current_zoning4  -7.344e+05  1.437e+05  -5.111 3.25e-07 ***
## current_zoning5  -7.237e+05  3.751e+05  -1.929 0.053711 .
## current_zoning6  -6.579e+05  1.369e+05  -4.806 1.56e-06 ***
## current_zoning7  -7.115e+05  1.364e+05  -5.215 1.87e-07 ***

```

```

## current_zoning8      -8.201e+05  1.570e+05  -5.223  1.79e-07 ***
## current_zoning9      -6.988e+05  1.883e+05  -3.711  0.000207 ***
## current_zoning10     -6.510e+05  1.355e+05  -4.804  1.58e-06 ***
## current_zoning11     -6.096e+05  1.345e+05  -4.533  5.86e-06 ***
## current_zoning12     -6.238e+05  1.400e+05  -4.455  8.47e-06 ***
## current_zoning13     -6.950e+05  1.344e+05  -5.170  2.38e-07 ***
## current_zoning14     -6.987e+05  1.345e+05  -5.195  2.08e-07 ***
## current_zoning15     -1.692e+05  1.510e+05  -1.120  0.262663
## current_zoning16     -2.335e+05  1.432e+05  -1.630  0.103021
## current_zoning17     -7.061e+05  1.408e+05  -5.015  5.37e-07 ***
## current_zoning18     -2.580e+05  2.457e+05  -1.050  0.293614
## current_zoning19     -6.412e+05  1.356e+05  -4.730  2.27e-06 ***
## current_zoning20     -7.213e+05  1.671e+05  -4.317  1.59e-05 ***
## current_zoning21     -6.596e+05  1.462e+05  -4.512  6.48e-06 ***
## current_zoning22     -7.120e+05  1.338e+05  -5.320  1.05e-07 ***
## current_zoning23     -7.741e+05  1.400e+05  -5.530  3.27e-08 ***
## current_zoning24     -6.495e+05  1.347e+05  -4.820  1.45e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 350000 on 12833 degrees of freedom
## Multiple R-squared:  0.2528, Adjusted R-squared:  0.251
## F-statistic: 140.1 on 31 and 12833 DF,  p-value: < 2.2e-16
#Multiple R-squared: 0.2528, Adjusted R-squared: 0.251
#F-statistic: 140.1 on 31 and 12833 DF, p-value: < 2.2e-16

# Run 3 : Recreating multiple regression model by applying backward elimination
# of non significant variables. current_zoning appears non-significant with
# some created variables of it having p-values > 0.05
hs_data_multi_rgr_mdl_3 <- lm(Sale_Price ~ sale_reason+building_grade+
square_feet_total_living+bedrooms+bath_full_count+bath_half_count+year_built+sq_ft_lot, data = mydata)
# check summary of multiple regression model
summary(hs_data_multi_rgr_mdl_3)

##
## Call:
## lm(formula = Sale_Price ~ sale_reason + building_grade + square_feet_total_living +
##     bedrooms + bath_full_count + bath_half_count + year_built +
##     sq_ft_lot, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2130758 -119409  -44728   41346  3722325
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -4.238e+06  4.339e+05  -9.767  < 2e-16 ***
## sale_reason    -1.163e+04  1.182e+03  -9.844  < 2e-16 ***
## building_grade  2.985e+04  4.446e+03   6.714  1.97e-11 ***
## square_feet_total_living 1.421e+02  5.957e+00  23.847  < 2e-16 ***
## bedrooms      -8.170e+03  4.593e+03  -1.779   0.0753 .
## bath_full_count  1.384e+04  6.073e+03   2.279   0.0227 *
## bath_half_count  5.876e+03  6.324e+03   0.929   0.3528
## year_built      2.159e+03  2.202e+02   9.806  < 2e-16 ***

```

```
## sq_ft_lot                2.972e-01  5.888e-02   5.046 4.56e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 355100 on 12856 degrees of freedom
## Multiple R-squared:  0.2295, Adjusted R-squared:  0.2291
## F-statistic: 478.8 on 8 and 12856 DF,  p-value: < 2.2e-16

# Intercept has drastically changed. Also p-values for bedrooms and
# bath_half_count have changed and they no more significant and
# have p-values > 0.05

# Run 4 : Recreating multiple regression model by applying backward elimination
# of non significant variables. bedrooms, bath_half_count appears non-significant
# with p-values > 0.05
hs_data_multi_rgr_mdl_4 <- lm(Sale_Price ~ sale_reason+building_grade+
square_feet_total_living+bath_full_count+year_built+sq_ft_lot, data = mydata)

# check summary of multiple regression model
summary(hs_data_multi_rgr_mdl_4)

##
## Call:
## lm(formula = Sale_Price ~ sale_reason + building_grade + square_feet_total_living +
##     bath_full_count + year_built + sq_ft_lot, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2136788 -119779  -44670   41079  3720748
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -4.477e+06  4.165e+05 -10.749  < 2e-16 ***
## sale_reason     -1.156e+04  1.180e+03  -9.792  < 2e-16 ***
## building_grade    3.106e+04  4.404e+03   7.053 1.84e-12 ***
## square_feet_total_living 1.374e+02  5.083e+00  27.038  < 2e-16 ***
## bath_full_count  1.288e+04  6.040e+03   2.132   0.033 *
## year_built       2.268e+03  2.132e+02  10.640  < 2e-16 ***
## sq_ft_lot        3.132e-01  5.805e-02   5.395 6.97e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 355100 on 12858 degrees of freedom
## Multiple R-squared:  0.2293, Adjusted R-squared:  0.2289
## F-statistic: 637.6 on 6 and 12858 DF,  p-value: < 2.2e-16

# Intercept seems to have stabilized with highly significant p-value
# All remaining predictor variables are highly significant with p-values < 0.05
# F-statistic has significantly improved with highly significant p-value
# Residual standard error: 355100 on 12858 degrees of freedom
# Multiple R-squared:  0.2293, Adjusted R-squared:  0.2289
# F-statistic: 637.6 on 6 and 12858 DF,  p-value: < 2.2e-16
```

Comparing simple regression to multi regression model

1. Multiple R-squared (variation explained) - In simple regression sq_ft_lot explains only 1.4% of variation in Sale_Price while in final multi regression model 4 predictors together explains almost 23% variation in Sale_Price.
2. Adjusted R-squared - Difference between adjusted R-squared and R-squared in multi regression model is 0.0004 while same difference in simple linear regression is 0.001 which is higher. In ideal world we would like adjusted R-squared to be same as R-squared to say that model can be more generalized and represent population.
3. P-value of predictors - P-values for each predictor in both simple regression and multi regression are highly significant and almost all of them are well below critical value of 0.05.
4. F-statistic - Fscore of multi regression model is almost 4.5 times that of simple regression model. P-value of both models is highly significant. This means that R-squared of multi regression model is more significant than that of simple regression model.

Overall multiple regression model is a much better predictive model for Sale_Price

b. iv. Considering the parameters of the multiple regression model you

have created. What are the standardized betas for each parameter and

what do the values indicate?

Absolute values of beta coefficients of multi regression model are comparatively higher and showing higher strength of the effect of each individual independent variable to the dependent variable. Most of the coefficients are positive showing positive effect on Sale_price.

This means that unit change in square_feet_total_living will cause Sale_Price to go up by 137.4 units, while keeping effect of all other predictors constant. We can say that effect of sq_feet_tital_living is more compared to sq_ft_lot on Sale_Price when then are seen individually keep other predictors constant. p-values associated with t values of predictors are very significant and well below cut-off value of 0.05 and in most case below 0.01 which would they will be significant in more than 99% cases. With t value of 27.038 and p-value < 0.01

square_feet_total_living is most significant predictor of the model.

```
summary(hs_data_multi_rgr_mdl_4)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sale_reason + building_grade + square_feet_total_living +
##     bath_full_count + year_built + sq_ft_lot, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2136788  -119779   -44670    41079   3720748
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -4.477e+06  4.165e+05 -10.749  < 2e-16 ***
## sale_reason    -1.156e+04  1.180e+03  -9.792  < 2e-16 ***
## building_grade   3.106e+04  4.404e+03   7.053 1.84e-12 ***
## square_feet_total_living 1.374e+02  5.083e+00  27.038  < 2e-16 ***
## bath_full_count  1.288e+04  6.040e+03   2.132   0.033 *
## year_built      2.268e+03  2.132e+02  10.640  < 2e-16 ***
## sq_ft_lot       3.132e-01  5.805e-02   5.395 6.97e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 355100 on 12858 degrees of freedom
## Multiple R-squared:  0.2293, Adjusted R-squared:  0.2289
## F-statistic: 637.6 on 6 and 12858 DF,  p-value: < 2.2e-16
```

Absolute value of beta coefficient of simple regression model is comparatively lower i.e. 0.85 and thus showing weaker strength of the effect of predictor variable to predicted variable. Coefficient is positive showing positive relationship. One unit change in sq_ft_lot will cause 0.85 units of change in Sale_Price. p-value for sq_ft_lot is highly significant and well below 0.01 which means in 99%+ cases this effect will stand true.

```
summary(hs_data_smpl_rgr_mdl)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sq_ft_lot, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2016064  -194842   -63293    91565   3735109
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 6.418e+05  3.800e+03  168.90  <2e-16 ***
## sq_ft_lot    8.510e-01  6.217e-02  13.69  <2e-16 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 401500 on 12863 degrees of freedom
## Multiple R-squared:  0.01435,    Adjusted R-squared:  0.01428
## F-statistic: 187.3 on 1 and 12863 DF,  p-value: < 2.2e-16
```

We can also get standardized beta estimates using `lm.beta()` of QuantPsyc package.

```
library(QuantPsyc)
```

```
## Loading required package: boot
## Loading required package: MASS
##
## Attaching package: 'MASS'
## The following object is masked from 'package:dplyr':
##
##      select
##
## Attaching package: 'QuantPsyc'
## The following object is masked from 'package:base':
##
##      norm
```

```
lm.beta(hs_data_multi_rgr_mdl_4)
```

```
##           sale_reason           building_grade square_feet_total_living
##           -0.07645294             0.08392564             0.33641067
##           bath_full_count           year_built             sq_ft_lot
##           0.02072611             0.09658661             0.04409583
```

As standardized beta values are calculated in standard deviation units for all predictors they can be easily compared to each other. We can see that for `square_feet_total_living` standard beta value is 0.33 (one standard deviation unit increase in `square_feet_total_living` will cause 0.33 standard deviation unit increase in Sale Price) which is much higher than any other predictor's value and thus it indicates highest effect of `square_feet_total_living` on `Sale_Price`.

b v. Calculate the confidence intervals for the parameters in your model and

explain what the results indicate.

```
# calculating confidence interval for the parameters of simple regression model
confint(hs_data_smpl_rgr_mdl)
```

```
##           2.5 %           97.5 %
## (Intercept) 6.343730e+05 6.492698e+05
## sq_ft_lot   7.291208e-01 9.728641e-01
```

```
# calculating confidence interval for the parameters of multiple regression model 4
confint(hs_data_multi_rgr_md1_4)
```

```
##                2.5 %      97.5 %
## (Intercept)    -5.293246e+06 -3.660485e+06
## sale_reason    -1.386989e+04 -9.243201e+03
## building_grade  2.242855e+04  3.969337e+04
## square_feet_total_living 1.274738e+02 1.474013e+02
## bath_full_count 1.038854e+03 2.471804e+04
## year_built     1.850308e+03 2.685978e+03
## sq_ft_lot      1.994069e-01 4.269936e-01
```

In simple linear regression model confidence interval is comparatively less wide and thus sample b value of predictor sq_ft_lot is more representative of being closer to beta value in population. In multi regression model, confidence intervals for few variables like bath_full_count, sale_reason, and even building_grade are wider and thus we are comparatively lesser confident about the mean of the future values. It is generally suggested to increase the sample size if CI is too wide, which doesn't seem to be the case in either model. In either model confidence interval does not cross zero for any of the predictors and thus it is indicative that relationships strength is good between predictors and predicted variables and thus model will not be predicting positive in some samples while negative in other samples because of particular predictor.

b. vi. Assess the improvement of the new model compared to your original model

(simple regression model) by testing whether this change is significant by

performing an analysis of variance.

```
# Running ANOVA for simple regression model
aov(Sale_Price ~ sq_ft_lot, data = mydata)
```

```
## Call:
##   aov(formula = Sale_Price ~ sq_ft_lot, data = mydata)
##
## Terms:
##              sq_ft_lot    Residuals
## Sum of Squares 3.019674e+13 2.073377e+15
## Deg. of Freedom      1      12863
##
## Residual standard error: 401483.8
## Estimated effects may be unbalanced
```

```

# Running ANOVA for multiple regression model
aov(Sale_Price ~ sale_reason+building_grade+
square_feet_total_living+bath_full_count+year_built+sq_ft_lot, data = mydata)

## Call:
##   aov(formula = Sale_Price ~ sale_reason + building_grade + square_feet_total_living +
##       bath_full_count + year_built + sq_ft_lot, data = mydata)
##
## Terms:
##               sale_reason building_grade square_feet_total_living
## Sum of Squares  2.862062e+13   3.089249e+14                1.249171e+14
## Deg. of Freedom           1                 1                     1
##               bath_full_count   year_built      sq_ft_lot   Residuals
## Sum of Squares    4.206324e+12  1.202280e+13  3.669902e+12  1.621212e+15
## Deg. of Freedom           1                 1                 1      12858
##
## Residual standard error: 355085.7
## Estimated effects may be unbalanced

```

We can observe that Residual standard error in case of simple linear regression is higher than that of multi regression model and thus there is more unexplained variation in data by simple regression model predictor that multi regression model predictors together. In other words multi regression model is better model to explain variation in the data.

b. vii. Perform casewise diagnostics to identify outliers and/or influential cases,

storing each function's output in a dataframe assigned to a unique variable name.

Statistics to identify outliers in the data by checking residuals using `resid()`, standard residuals using `rstandard()` and studentized residual using `rstudent()`.

Identifying influential cases -

1. Cook's distance - Access overall influence of a single case on the model.

Cook and Weisberg (1982) have suggested that values greater than 1 may be cause for concern. If a point is a significant outlier on Y, but its Cook's distance is < 1 , there is no real need to delete that point since it does not have a large effect on the regression analysis. Using `cooks.distance()`

2. Hat value (leverage) - Value of 0 would mean that case has no influence, while value of 1 would indicate that case has complete influence over prediction. If no cases exert undue influence over the model then we would expect all of the leverage values to be close to the average value $((k + 1)/n)$. Using `hatvalues()`.

3. DFFit & DFbeta - It is the difference between the predicted value for a

case when the model is fitted including that case and when the model is fitted excluding that case. Using `dffits()` and `DFBeta` using `dfbeta()`.
 adding variables to check each of above case level statistics to original data frames.

```
# adding case level residual/outlier and influential stats for multi regression model
mydata$residuals <- resid(hs_data_multi_rgr_mdl_4)
mydata$standardized.residuals <- rstandard(hs_data_multi_rgr_mdl_4)
mydata$studentized.residuals <- rstudent(hs_data_multi_rgr_mdl_4)
mydata$cooks.distance <- cooks.distance(hs_data_multi_rgr_mdl_4)
mydata$dfbeta <- dfbeta(hs_data_multi_rgr_mdl_4)
mydata$dffit <- dffits(hs_data_multi_rgr_mdl_4)
mydata$leverage <- hatvalues(hs_data_multi_rgr_mdl_4)
mydata$covariance.ratios <- covratio(hs_data_multi_rgr_mdl_4)
# looking at the data
# mydata
# writing the saved stats for each case into a table
write.table(mydata, "House Sales With Diagnostics.dat", sep = "\t", row.names = FALSE)
# nrow(mydata) -- 12865
# check if about 5% of cases (<= 643 cases) have standardized residual within +-2.
sum(mydata$standardized.residuals > 2 | mydata$standardized.residuals < -2)
```

```
## [1] 329
```

```
# As only 329 cases are outside range or have large residuals, we are well within the range of 5% outliers
# To exactly identify outliers we can add a variable called large.residual in the data frame to save the results
mydata$large.residual <- mydata$standardized.residuals > 2 | mydata$standardized.residuals < -2
# we can now select the outlier cases by select rows with large.residual = TRUE
# model is fairly accurate

# check how many cases have standard residuals > 3 which may be we can investigate further
sum(mydata$standardized.residuals > 3 | mydata$standardized.residuals < -3)
```

```
## [1] 213
```

```
# create a variable to flag cases with very large residual
mydata$very.large.residual <- mydata$standardized.residuals > 3 | mydata$standardized.residuals < -3
# 213 cases out of 12865 -- about 1.65%

# Let's look at the leverage (hat value), cook's distance, and covariance ratio
# for cases with large.residual = TRUE
mydata[mydata$large.residual, c("cooks.distance", "leverage", "covariance.ratios")]
```

##	cooks.distance	leverage	covariance.ratios
## 25	0.0010257923	0.0011869228	0.9984429
## 108	0.0033044011	0.0042136523	1.0017920
## 115	0.0024117590	0.0018063670	0.9972755
## 178	0.0011465389	0.0012854582	0.9984365
## 239	0.0001090031	0.0001668056	0.9982225
## 246	0.0031564325	0.0017464582	0.9954241
## 287	0.0006893490	0.0007933254	0.9980305
## 295	0.2396277578	0.1188300220	1.1278061
## 300	0.0075267399	0.0043854071	0.9984257

## 341	0.0004487567	0.0006708699	0.9986685
## 344	0.0057351762	0.0089305205	1.0071144
## 359	0.0017397504	0.0014731451	0.9975264
## 385	0.0025307845	0.0015599477	0.9959382
## 396	0.0014805940	0.0014632775	0.9981591
## 475	0.0006986847	0.0009934765	0.9988611
## 482	0.0031738517	0.0039360295	1.0014278
## 508	0.0033052073	0.0053950353	1.0036382
## 528	0.0010875245	0.0016248737	0.9996238
## 576	0.0008319101	0.0013155085	0.9994538
## 661	0.0027968265	0.0028437415	0.9996540
## 670	0.0016660191	0.0005997724	0.9906016
## 679	0.0070450303	0.0036509393	0.9968763
## 742	0.0004484261	0.0007724219	0.9991065
## 784	0.0019524886	0.0023320386	0.9996953
## 802	0.0010311108	0.0015000453	0.9994297
## 811	0.0035835198	0.0009199438	0.9867072
## 852	0.0009102800	0.0015231355	0.9997943
## 853	0.0004455858	0.0007137136	0.9988811
## 877	0.0016204219	0.0014373881	0.9976943
## 916	0.0045527934	0.0030400158	0.9978991
## 1009	0.0002877654	0.0004295533	0.9984230
## 1099	0.0009422438	0.0015341461	0.9997424
## 1119	0.0020112581	0.0015651144	0.9972234
## 1142	0.0011810063	0.0010314525	0.9972200
## 1155	0.0149292319	0.0095001048	1.0041643
## 1305	0.0011346071	0.0016966438	0.9996980
## 1345	0.0002542520	0.0004161686	0.9986340
## 1368	0.0007075141	0.0011555917	0.9993700
## 1380	0.0003101548	0.0004637808	0.9984617
## 1442	0.0016275507	0.0015502173	0.9981018
## 1492	0.0010091685	0.0012623284	0.9987649
## 1504	0.0048822856	0.0030902960	0.9976377
## 1543	0.0010641713	0.0017134567	0.9998962
## 1550	0.0008218085	0.0008187770	0.9975437
## 1633	0.0001621965	0.0002170445	0.9979159
## 1650	0.0135138750	0.0062910557	0.9987168
## 1716	0.0009970945	0.0007971874	0.9965835
## 1745	0.0009432943	0.0007656760	0.9966233
## 1870	0.0074126080	0.0066798036	1.0030493
## 1962	0.0024740915	0.0014260536	0.9953770
## 1963	0.0006106903	0.0007478570	0.9981840
## 1964	0.0008404656	0.0004465255	0.9938367
## 1976	0.0013242658	0.0003997737	0.9883823
## 1977	0.0006853792	0.0001884214	0.9869458
## 1978	0.0012447634	0.0003547031	0.9875953
## 1979	0.0006774109	0.0001869084	0.9869939
## 1980	0.0012740900	0.0003552299	0.9873042
## 1981	0.0014775095	0.0004630181	0.9889040
## 1982	0.0012863954	0.0003793806	0.9880669
## 2022	0.0013424594	0.0009537856	0.9961454
## 2099	0.0005363480	0.0006876495	0.9982629
## 2137	0.0005589791	0.0003740237	0.9952344
## 2157	0.0005565648	0.0005217858	0.9970069

## 2264	0.0031592360	0.0015138722	0.9941316
## 2302	0.0017626301	0.0024168498	1.0001911
## 2360	0.0005459138	0.0005608369	0.9974005
## 2361	0.0024410539	0.0009953756	0.9922280
## 2469	0.0004473775	0.0003714912	0.9963336
## 2684	0.0015598153	0.0004735320	0.9885261
## 2685	0.0012316856	0.0003774574	0.9885464
## 2686	0.0014329427	0.0005141746	0.9904813
## 2687	0.0012141040	0.0003613572	0.9881655
## 2688	0.0011725802	0.0003451790	0.9880091
## 2689	0.0013120914	0.0004579103	0.9901284
## 2690	0.0013092871	0.0004699203	0.9904396
## 2699	0.0340776069	0.0254130949	1.0215325
## 2708	0.0013991091	0.0008301737	0.9949675
## 2709	0.0012136735	0.0008456224	0.9959317
## 2710	0.0017277029	0.0010943362	0.9956365
## 2717	0.0030168291	0.0015203003	0.9945262
## 2742	0.0005096100	0.0004644780	0.9968336
## 2852	0.0323551464	0.0084297003	0.9945063
## 2934	0.0009961450	0.0007971874	0.9965880
## 2937	0.0009613303	0.0007971874	0.9967538
## 3097	0.0026676183	0.0014175848	0.9948117
## 3102	0.0007357039	0.0006850037	0.9971425
## 3110	0.0008237369	0.0008555478	0.9977359
## 3111	0.0003968700	0.0005785581	0.9985112
## 3168	0.0039875726	0.0004821071	0.9699125
## 3169	0.0033779971	0.0004123887	0.9701391
## 3170	0.0029576959	0.0003738115	0.9711367
## 3171	0.0027292290	0.0003492111	0.9714720
## 3172	0.0027537800	0.0003513317	0.9713897
## 3173	0.0027317318	0.0003496169	0.9714795
## 3174	0.0027533928	0.0003512709	0.9713887
## 3175	0.0030548353	0.0003796849	0.9706469
## 3176	0.0027749988	0.0003587433	0.9717792
## 3177	0.0026791022	0.0003464883	0.9717788
## 3178	0.0026799098	0.0003466282	0.9717818
## 3179	0.0028631251	0.0003611646	0.9710675
## 3180	0.0026810696	0.0003468243	0.9717859
## 3181	0.0028623150	0.0003610450	0.9710660
## 3182	0.0039884401	0.0004822319	0.9699139
## 3183	0.0039868755	0.0004820066	0.9699114
## 3184	0.0039868755	0.0004820066	0.9699114
## 3185	0.0039885428	0.0004822466	0.9699141
## 3186	0.0039875726	0.0004821071	0.9699125
## 3187	0.0039875726	0.0004821071	0.9699125
## 3188	0.0033853118	0.0004134073	0.9701493
## 3189	0.0030536527	0.0003795151	0.9706449
## 3190	0.0027516005	0.0003509865	0.9713838
## 3191	0.0030531822	0.0003794474	0.9706441
## 3192	0.0027294619	0.0003492496	0.9714728
## 3193	0.0027512095	0.0003509237	0.9713827
## 3194	0.0027509545	0.0003508825	0.9713819
## 3195	0.0026794947	0.0003465567	0.9717803
## 3196	0.0026803440	0.0003467022	0.9717834

## 3197	0.0027812390	0.0003597503	0.9717962
## 3198	0.0039902426	0.0004824902	0.9699168
## 3199	0.0039900914	0.0004824686	0.9699165
## 3200	0.0039868755	0.0004820066	0.9699114
## 3201	0.0039868755	0.0004820066	0.9699114
## 3202	0.0039868755	0.0004820066	0.9699114
## 3424	0.0024495202	0.0040292686	1.0022766
## 3464	0.0029958323	0.0003794015	0.9712020
## 3465	0.0028920607	0.0003723198	0.9716726
## 3466	0.0026276922	0.0003464232	0.9723251
## 3467	0.0026294974	0.0003467421	0.9723320
## 3468	0.0029970621	0.0003795821	0.9712041
## 3469	0.0028044725	0.0003604398	0.9716123
## 3470	0.0026272739	0.0003463465	0.9723233
## 3471	0.0028929496	0.0003724558	0.9716744
## 3472	0.0026776797	0.0003492414	0.9720233
## 3473	0.0027005098	0.0003511464	0.9719381
## 3474	0.0029999885	0.0003800094	0.9712089
## 3475	0.0026801737	0.0003496534	0.9720308
## 3476	0.0037772871	0.0004924125	0.9721523
## 3477	0.0026793022	0.0003495111	0.9720283
## 3478	0.0039134029	0.0004820066	0.9704772
## 3479	0.0039134029	0.0004820066	0.9704772
## 3480	0.0039134029	0.0004820066	0.9704772
## 3481	0.0039134029	0.0004820066	0.9704772
## 3482	0.0039134029	0.0004820066	0.9704772
## 3483	0.0039134076	0.0004820073	0.9704772
## 3484	0.0033146796	0.0004122789	0.9707011
## 3485	0.0028051885	0.0003710987	0.9724468
## 3486	0.0028040584	0.0003603760	0.9716114
## 3487	0.0027138345	0.0003583941	0.9723855
## 3488	0.0026989172	0.0003508860	0.9719335
## 3489	0.0026767519	0.0003490836	0.9720201
## 3490	0.0027005680	0.0003511558	0.9719383
## 3491	0.0037826809	0.0004935759	0.9721801
## 3492	0.0028922054	0.0003723420	0.9716729
## 3493	0.0026280542	0.0003464886	0.9723266
## 3494	0.0029973689	0.0003796270	0.9712046
## 3495	0.0039134076	0.0004820073	0.9704772
## 3496	0.0039134076	0.0004820073	0.9704772
## 3497	0.0039134076	0.0004820073	0.9704772
## 3523	0.0026507975	0.0015849983	0.9957735
## 3837	0.0006838252	0.0007383763	0.9977580
## 3918	0.0012625486	0.0015651144	0.9990412
## 3919	0.0007652759	0.0010314525	0.9987521
## 4055	0.0014804592	0.0011854653	0.9969800
## 4056	0.0045005009	0.0013619354	0.9893770
## 4248	0.0007189969	0.0012086223	0.9994892
## 4391	0.0011440636	0.0019973401	1.0003650
## 4435	0.0003501356	0.0004602009	0.9981078
## 4648	0.1052345743	0.0102109306	0.9722188
## 4649	0.6906373336	0.0441092700	0.9884534
## 4671	0.0033942707	0.0042189274	1.0017201
## 4695	0.0015209988	0.0004222043	0.9873116

## 4696	0.0012415938	0.0008215703	0.9956192
## 4740	0.0003991572	0.0006797623	0.9989884
## 4821	0.0018888427	0.0015127881	0.9973094
## 4834	0.0011454391	0.0009537856	0.9969297
## 4840	0.0156021509	0.0095001048	1.0038956
## 4934	0.0009536065	0.0009580187	0.9977148
## 5083	0.0901980470	0.0309037562	1.0213738
## 5491	0.0001204817	0.0001827619	0.9982166
## 5494	0.0001184603	0.0001828554	0.9982600
## 5495	0.0001210330	0.0001833950	0.9982144
## 5496	0.0001205418	0.0001828299	0.9982163
## 5497	0.0002495714	0.0003894789	0.9984937
## 5549	0.0007267978	0.0011075487	0.9991545
## 5935	0.0013552866	0.0010634637	0.9967606
## 6055	0.0009548636	0.0013018514	0.9990555
## 6230	0.0016927822	0.0003674489	0.9834783
## 6231	0.0016557276	0.0003716195	0.9840513
## 6232	0.0018284484	0.0004261427	0.9847252
## 6233	0.0018688703	0.0003973082	0.9831440
## 6234	0.0016795075	0.0003655666	0.9835237
## 6235	0.0016921418	0.0003673263	0.9834790
## 6236	0.0016791110	0.0003654903	0.9835241
## 6237	0.0017228537	0.0003748257	0.9835249
## 6238	0.0019715318	0.0004140669	0.9829447
## 6239	0.0016560923	0.0003716927	0.9840510
## 6429	0.0051922484	0.0003579187	0.9468824
## 6430	0.0028656656	0.0001849280	0.9431183
## 6431	0.0034505351	0.0002283094	0.9445494
## 6432	0.0040677930	0.0002746190	0.9456881
## 6433	0.0044944781	0.0003068287	0.9463174
## 6434	0.0048150696	0.0003295367	0.9464729
## 6435	0.0087269903	0.0006216813	0.9488530
## 6436	0.0035389281	0.0002355183	0.9448734
## 6437	0.0028910251	0.0001879459	0.9435341
## 6438	0.0030345974	0.0001959032	0.9431503
## 6439	0.0044687097	0.0003045868	0.9462308
## 6440	0.0050615201	0.0003607200	0.9486138
## 6441	0.0072029810	0.0004873128	0.9460152
## 6442	0.0044465560	0.0003022794	0.9460878
## 6443	0.0062548245	0.0004833333	0.9527154
## 6444	0.0065752820	0.0005214943	0.9539694
## 6445	0.0066645597	0.0004997116	0.9512863
## 6446	0.0063454767	0.0004937736	0.9530550
## 6447	0.0071497350	0.0005789957	0.9549787
## 6448	0.0026806822	0.0002002402	0.9508024
## 6449	0.0059395065	0.0004822235	0.9549975
## 6450	0.0077410779	0.0006442085	0.9562607
## 6451	0.0065804276	0.0004930996	0.9512497
## 6452	0.0063582774	0.0004955632	0.9531320
## 6453	0.0076861191	0.0006311932	0.9556587
## 6454	0.0067255502	0.0005369915	0.9542924
## 6455	0.0062439332	0.0004799401	0.9524606
## 6456	0.0029151760	0.0002198457	0.9512864
## 6457	0.0047098882	0.0003741964	0.9539020

## 6512	0.0027687042	0.0013074985	0.9938082
## 6527	0.0011439540	0.0014878729	0.9991076
## 6739	0.0010574400	0.0010579402	0.9977993
## 6796	0.0002679505	0.0004499813	0.9987267
## 6938	0.0008439725	0.0003966984	0.9928578
## 6939	0.0012674214	0.0006937685	0.9942942
## 6940	0.0005499315	0.0002296020	0.9916774
## 6941	0.0005484610	0.0002296106	0.9917020
## 6942	0.0011257892	0.0005881841	0.9938582
## 6943	0.0021717622	0.0011535140	0.9945433
## 6944	0.0015445694	0.0008612923	0.9945893
## 6945	0.0008439725	0.0003966984	0.9928578
## 6946	0.0005520202	0.0002309316	0.9916967
## 6947	0.0005282048	0.0002164612	0.9914939
## 6948	0.0005193341	0.0002109403	0.9914059
## 7039	0.0008794222	0.0013133248	0.9993099
## 7147	0.0008071568	0.0008465522	0.9977625
## 7167	0.0040556743	0.0007040698	0.9794912
## 7210	0.0034382281	0.0013317753	0.9920764
## 7211	0.0049454782	0.0015271078	0.9897918
## 7446	0.0019642756	0.0002300190	0.9686643
## 7447	0.0019171957	0.0002216273	0.9682447
## 7448	0.0019661194	0.0002303729	0.9686836
## 7449	0.0019179172	0.0002216866	0.9682413
## 7450	0.0019172881	0.0002215034	0.9682251
## 7451	0.0019167196	0.0002213281	0.9682090
## 7452	0.0019168567	0.0002216008	0.9682465
## 7453	0.0018489967	0.0002157551	0.9685378
## 7454	0.0018541697	0.0002165786	0.9685709
## 7455	0.0018138062	0.0002095564	0.9682144
## 7456	0.0019172478	0.0002216315	0.9682445
## 7457	0.0019590073	0.0002291654	0.9686307
## 7458	0.0019156108	0.0002212414	0.9682150
## 7459	0.0019175169	0.0002199960	0.9679999
## 7460	0.0019172881	0.0002215034	0.9682251
## 7461	0.0019152975	0.0002203961	0.9680963
## 7462	0.0019160039	0.0002201497	0.9680481
## 7463	0.0019152776	0.0002203946	0.9680964
## 7507	0.0023158182	0.0024123317	0.9993094
## 7649	0.0006917948	0.0011398730	0.9993746
## 7650	0.0006918818	0.0011402519	0.9993754
## 7683	0.0103781570	0.0159671542	1.0143040
## 7791	0.0010461800	0.0015224008	0.9994529
## 7871	0.0046273592	0.0030157643	0.9977352
## 8154	0.0002961755	0.0004439867	0.9984482
## 8232	0.0024292806	0.0023320386	0.9989182
## 8262	0.0253133456	0.0181771335	1.0137703
## 8320	0.0064410706	0.0050596919	1.0007889
## 8377	0.3807445906	0.0644221375	1.0471077
## 8457	0.0005906999	0.0007478570	0.9982856
## 8458	0.0008222984	0.0004465255	0.9939909
## 8535	0.0003569741	0.0005152238	0.9984215
## 8541	0.0001631968	0.0002170445	0.9978983
## 8698	0.0002796865	0.0003719032	0.9980529

```
## 8710    0.0022473554 0.0007041959    0.9891443
## 8763    0.0005216706 0.0006359458    0.9980577
## 8887    0.0129194672 0.0027836422    0.9857741
## 8911    0.0013009607 0.0019755686    1.0000169
## 8946    0.0013009607 0.0019755686    1.0000169
## 9215    0.0024188561 0.0026122144    0.9996397
## 9293    0.0011160547 0.0016248737    0.9995570
## 9369    0.0013616896 0.0021917458    1.0003758
## 9420    0.0060749032 0.0073714500    1.0048367
## 9453    0.0005106974 0.0007971874    0.9989027
## 9528    0.0056161190 0.0022635893    0.9933926
## 9546    0.0006538835 0.0007383763    0.9979122
## 9722    0.0007277701 0.0006797623    0.9971501
## 10125   0.0052324495 0.0048478089    1.0013104
## 10318   0.0133567782 0.0224122117    1.0212129
## 10371   0.0021064306 0.0028288247    1.0005472
## 10418   0.0015299893 0.0023617801    1.0004458
## 10478   0.0004579563 0.0007820097    0.9990968
## 10623   0.0004269388 0.0006254808    0.9985711
## 10707   0.0001299349 0.0002014687    0.9982897
## 10723   0.0004312082 0.0006878271    0.9988453
## 10741   0.0002559881 0.0004202566    0.9986450
## 10787   0.0027445767 0.0014033335    0.9945172
## 10844   0.0016949044 0.0021391104    0.9996722
## 10958   0.0004796424 0.0004712638    0.9971420
## 10995   0.0030409613 0.0051210076    1.0034327
## 11165   0.0013206888 0.0014774990    0.9986218
## 11289   0.0006334948 0.0002389100    0.9907170
## 11413   0.0027417091 0.0012059380    0.9931160
## 11558   0.0042468752 0.0009956524    0.9853908
## 11586   0.0016574128 0.0011102733    0.9959786
## 11728   0.0068990834 0.0047846693    0.9998698
## 11758   0.0022656737 0.0028288247    1.0003331
## 11772   0.0120241217 0.0018168689    0.9774022
## 11822   0.0025447045 0.0006655231    0.9867224
## 11898   0.0327527392 0.0348098650    1.0330472
## 11899   0.0767419624 0.0375479305    1.0318106
## 11982   0.0086379247 0.0030400158    0.9928114
## 11992   0.0732808662 0.0046287603    0.9463640
## 12212   0.0020701946 0.0019228612    0.9983743
## 12255   0.0041240793 0.0022774808    0.9959444
## 12256   0.0011712201 0.0011821940    0.9979574
## 12392   0.0008750702 0.0010714059    0.9985077
## 12472   0.0042021297 0.0013192567    0.9897842
## 12487   0.0110948252 0.0046009420    0.9960107
## 12577   0.0039696302 0.0014696028    0.9917631
## 12582   0.0186385515 0.0024805710    0.9747313
## 12643   0.0160063267 0.0019543570    0.9716895
## 12686   0.0001473875 0.0002401013    0.9984466
## 12764   0.0054974616 0.0021949561    0.9932346
```

```
# check if any outlier cases have cook's distance > 1
sum(mydata[mydata$large.residual, c("cooks.distance")] > 1)
```

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

```

# None of the cases have cooks distance > 1, so none of the cases have undue
# influence on the model

# calculate average leverage using formula = (k+1/n)
avg_leverage <- (6+1)/12865
# three times average leverage
times_3_leverage <- avg_leverage*3
# check if there are outlier cases with leverage > 3 times the average leverage
sum(mydata[mydata$large.residual, c("levarage")] > times_3_leverage)

## [1] 0
# There are none

```

b. viii. Calculate the standardized residuals using the appropriate command,

specifying those that are ± 2 , storing the results of large residuals in a

variable you create.

Answered this question as part of analysis in last question. Below is the command used. Variable is already added to the data frame mydata

```

- mydata$large.residual <- mydata$standardized.residuals > 2 | mydata$standardized.residuals <
-2 -

```

b. ix. Use the appropriate function to show the sum of large residuals.

Answered in b. vii. as part of analysis of residual and influence.

```

# nrow(mydata) -- 12865
# check if about 5% of cases (<= 643 cases) have standardized residual within +-2.
sum(mydata$standardized.residuals > 2 | mydata$standardized.residuals < -2)

## [1] 329
# As only 329 cases are outside range or have large residuals, we are well
# within the range of 5% outliers.

```

b. x. Which specific variables have large residuals (only cases that evaluate

as TRUE)?

```

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"square_feet_total_living","bedrooms",
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##	108		3230 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0972	47.63926
##	115		19656 NE REDMOND RD	98053	<NA>	-122.0772	47.69595
##	178		13414 WOODINVILLE REDMOND RD NE	98052	<NA>	-122.1343	47.72058
##	239		24103 NE 122ND ST	98053	<NA>	-122.0192	47.70719
##	246		3068 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0953	47.63820
##	287		28218 NE 40TH ST	98053	<NA>	-121.9611	47.64355
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##	344		1520 268TH AVE NE	98053	<NA>	-121.9787	47.62110
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##	385		20424 NE 64TH PL	98053	<NA>	-122.0685	47.66419
##	396		8024 255TH AVE NE	98053	<NA>	-121.9978	47.67461
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##	482		3624 175TH CT NE	98052	REDMOND	-122.1056	47.64296
##	508		7717 252ND AVE NE	98053	<NA>	-122.0033	47.67279
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##	576		25525 NE 67TH PL	98053	<NA>	-121.9982	47.66396
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##	679		24300 NE UNION HILL RD	98053	<NA>	-122.0140	47.66268
##	742		6920 242ND AVE NE	98053	<NA>	-122.0151	47.66757
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##	802		4402 244TH AVE NE	98053	<NA>	-122.0131	47.64924
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##	853		24629 NE 109TH ST	98053	<NA>	-122.0092	47.69466
##	877		8303 250TH AVE NE	98053	<NA>	-122.0064	47.67672
##	916		7350 259TH PL NE	98053	<NA>	-121.9927	47.66975
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##	1099		19736 NE 61ST PL	98053	<NA>	-122.0757	47.66093
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##	1345		16618 NE 47TH ST	98052	REDMOND	-122.1184	47.65191
##	1368		27935 E MAIN ST	98053	<NA>	-121.9663	47.60549
##	1380		3605 289TH AVE NE	98053	<NA>	-121.9524	47.64088
##	1442		3122 W AMES LAKE DR NE	98053	<NA>	-121.9623	47.63550
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## 1650	14338 WOODINVILLE REDMOND RD NE 98052	<NA> -122.1381 47.73097
## 1716	22340 NE 65TH PL 98053	<NA> -122.0407 47.66467
## 1745	23812 NE 61ST ST 98053	<NA> -122.0212 47.66215
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## 1962	8407 255TH AVE NE 98053	<NA> -122.0000 47.67678
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## 1964	2030 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0877 47.62828
## 1976	14026 NE 85TH CT 98052	REDMOND -122.1526 47.67923
## 1977	14009 NE 85TH CT 98052	REDMOND -122.1528 47.67900
## 1978	14007 NE 85TH CT 98052	REDMOND -122.1532 47.67900
## 1979	14005 NE 85TH CT 98052	REDMOND -122.1532 47.67926
## 1980	14006 NE 85TH CT 98052	REDMOND -122.1532 47.67940
## 1981	14018 NE 85TH CT 98052	REDMOND -122.1527 47.67959
## 1982	14022 NE 85TH CT 98052	REDMOND -122.1527 47.67941
## 2022	2140 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0874 47.62941
## 2099	5838 246TH PL NE 98053	<NA> -122.0097 47.65938
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## 2157	2240 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0875 47.63050
## 2264	2632 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0893 47.63440
## 2302	18446 NE 95TH ST 98052	REDMOND -122.0926 47.68585
## 2360	2350 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0878 47.63100
## 2361	2350 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0878 47.63100
## 2469	2768 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0914 47.63484
## 2684	5215 157TH CT NE 98052	REDMOND -122.1315 47.65483
## 2685	5221 157TH CT NE 98052	REDMOND -122.1315 47.65503
## 2686	5216 157TH CT NE 98052	REDMOND -122.1311 47.65488
## 2687	5209 157TH CT NE 98052	REDMOND -122.1315 47.65469
## 2688	5103 157TH CT NE 98052	REDMOND -122.1315 47.65451
## 2689	5210 157TH CT NE 98052	REDMOND -122.1311 47.65473
## 2690	5104 157TH CT NE 98052	REDMOND -122.1311 47.65456
## 2699	6415 196TH AVE NE 98053	<NA> -122.0798 47.66254
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## 2742	5826 249TH CT NE 98053	<NA> -122.0057 47.65873
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## 3110	2524 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0886 47.63300
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## 3464	11044 236TH PL NE 98053	<NA> -122.0241 47.69626
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## 3837	22465 NE 111TH LN 98053	<NA> -122.0384 47.69732
## 3918	7515 238TH AVE NE 98053	<NA> -122.0238 47.67172
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## 4056	2844 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0925 47.63591
## 4248	5106 272ND AVE NE 98053	<NA> -121.9751 47.65169
## 4391	5920 216TH PL NE 98053	<NA> -122.0503 47.66073
## 4435	15808 NE 51ST ST 98052	REDMOND -122.1297 47.65474
## 4648	12025 154TH PL NE 98052	<NA> -122.1350 47.70801
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## 4671	6910 264TH AVE NE 98053	<NA> -121.9852 47.66578
## 4695	2424 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0880 47.63192
## 4696	2424 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0880 47.63192
## 4740	2850 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0927 47.63610
## 4821	8436 143RD CT NE 98052	REDMOND -122.1489 47.67850
## 4834	2140 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0874 47.62941
## 4840	8700 196TH AVE NE 98053	<NA> -122.0782 47.67993
## 4934	3402 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0994 47.64049
## 5083	19805 NE NOVELTY HILL RD 98053	<NA> -122.0741 47.68643
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## 5494	16525 NE 43RD CT 98052	REDMOND -122.1193 47.64811
## 5495	16631 NE 43RD CT 98052	REDMOND -122.1191 47.64824
## 5496	16639 NE 43RD CT 98052	REDMOND -122.1187 47.64828
## 5497	16501 NE 43RD CT 98052	REDMOND -122.1199 47.64817
## 5549	8036 196TH AVE NE 98053	<NA> -122.0770 47.67611
## 5935	10100 203RD AVE NE 98053	<NA> -122.0682 47.68972
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## 6230	17144 NE 46TH CT 98052	REDMOND -122.1106 47.65080
## 6231	17372 NE 46TH CT 98052	REDMOND -122.1092 47.65088
## 6232	17376 NE 46TH CT 98052	REDMOND -122.1088 47.65092
## 6233	17380 NE 46TH CT 98052	REDMOND -122.1088 47.65078
## 6234	17252 NE 46TH CT 98052	REDMOND -122.1102 47.65079
## 6235	17256 NE 46TH CT 98052	REDMOND -122.1100 47.65079
## 6236	17260 NE 46TH CT 98052	REDMOND -122.1098 47.65079
## 6237	17364 NE 46TH CT 98052	REDMOND -122.1096 47.65078
## 6238	17368 NE 46TH CT 98052	REDMOND -122.1093 47.65083
## 6239	17384 NE 46TH CT 98052	REDMOND -122.1088 47.65062
## 6429	17137 NE 120TH ST 98052	REDMOND -122.1113 47.70674
## 6430	11818 171ST PL NE 98052	REDMOND -122.1119 47.70639
## 6431	17011 NE 118TH WAY 98052	REDMOND -122.1134 47.70580
## 6432	16943 NE 118TH WAY 98052	REDMOND -122.1138 47.70579
## 6433	16944 NE 118TH WAY 98052	REDMOND -122.1138 47.70624
## 6434	16909 NE 120TH ST 98052	REDMOND -122.1145 47.70694
## 6435	17128 NE 120TH ST 98052	REDMOND -122.1115 47.70718
## 6436	17136 NE 120TH ST 98052	REDMOND -122.1112 47.70716

## 6437	11902 171ST PL NE 98052	REDMOND -122.1120 47.70651
## 6438	11719 171ST PL NE 98052	REDMOND -122.1125 47.70568
## 6439	16955 NE 118TH WAY 98052	REDMOND -122.1135 47.70579
## 6440	11703 169TH PL NE 98052	REDMOND -122.1146 47.70573
## 6441	16906 NE 118TH WAY 98052	REDMOND -122.1146 47.70631
## 6442	17020 NE 118TH WAY 98052	REDMOND -122.1132 47.70624
## 6443	12212 164TH CT NE 98052	REDMOND -122.1203 47.70956
## 6444	12238 164TH CT NE 98052	REDMOND -122.1203 47.70980
## 6445	12300 164TH CT NE 98052	REDMOND -122.1203 47.71005
## 6446	12308 164TH CT NE 98052	REDMOND -122.1202 47.71018
## 6447	12351 164TH CT NE 98052	REDMOND -122.1212 47.71034
## 6448	12261 164TH CT NE 98052	REDMOND -122.1213 47.70954
## 6449	12249 164TH CT NE 98052	REDMOND -122.1208 47.70993
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## 6454	12375 164TH CT NE 98052	REDMOND -122.1212 47.71075
## 6455	12285 164TH CT NE 98052	REDMOND -122.1214 47.70982
## 6456	12273 164TH CT NE 98052	REDMOND -122.1214 47.70968
## 6457	12225 164TH CT NE 98052	REDMOND -122.1208 47.70963
## 6512	3018 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.0937 47.63761
## 6527	5120 159TH AVE NE 98052	REDMOND -122.1281 47.65495
## 6739	21117 NE 78TH ST 98053	<NA> -122.0580 47.67346
## 6796	17340 NE 36TH ST 98052	REDMOND -122.1078 47.64256
## 6938	11516 174TH CT NE 98052	REDMOND -122.1068 47.70294
## 6939	11527 174TH CT NE 98052	REDMOND -122.1078 47.70295
## 6940	17317 NE 116TH CT 98052	REDMOND -122.1079 47.70435
## 6941	17321 NE 116TH CT 98052	REDMOND -122.1080 47.70417
## 6942	11586 174TH CT NE 98052	REDMOND -122.1070 47.70340
## 6943	11504 174TH CT NE 98052	REDMOND -122.1070 47.70294
## 6944	11503 174TH CT NE 98052	REDMOND -122.1073 47.70294
## 6945	11519 174TH CT NE 98052	REDMOND -122.1075 47.70294
## 6946	11571 174TH CT NE 98052	REDMOND -122.1076 47.70324
## 6947	17327 NE 116TH CT 98052	REDMOND -122.1077 47.70417
## 6948	17333 NE 116TH CT 98052	REDMOND -122.1074 47.70418
## 7039	18852 NE 116TH ST 98052	<NA> -122.0875 47.70434
## 7147	6226 148TH AVE NE 98052	REDMOND -122.1428 47.66317
## 7167	3408 W LAKE SAMMAMISH PKWY NE 98052	REDMOND -122.1001 47.64084
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## 7649	13521 NE 70TH ST	98052	REDMOND	-122.1588	47.66749
## 7650	13524 NE 70TH ST	98052	REDMOND	-122.1588	47.66770
## 7683	2222 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0874	47.62981
## 7791	26510 NE 70TH ST	98053	<NA>	-121.9864	47.66943
## 7871	25025 NE 80TH ST	98053	<NA>	-122.0051	47.67247
## 8154	21817 NE 97TH PL	98053	<NA>	-122.0484	47.68620
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## 8262	26408 NE 70TH ST	98053	<NA>	-121.9857	47.66751
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## 8377	20210 NE 85TH ST	98053	<NA>	-122.0724	47.68034
## 8457	2030 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0877	47.62828
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## 8535	13304 NE 75TH ST	98052	REDMOND	-122.1624	47.67189
## 8541	8490 138TH LN NE	98052	REDMOND	-122.1554	47.67857
## 8698	3040 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0950	47.63783
## 8710	3244 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0973	47.63979
## 8763	2608 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0886	47.63358
## 8887	3026 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0947	47.63759
## 8911	2006 250TH PL NE	98074	SAMMAMISH	-122.0045	47.62730
## 8946	2006 250TH PL NE	98074	SAMMAMISH	-122.0045	47.62730
## 9215	12210 WILLOWS RD NE	98052	<NA>	-122.1536	47.70956
## 9293	2829 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0929	47.63521
## 9369	2722 288TH AVE NE	98053	<NA>	-121.9538	47.63305
## 9420	22614 NE UNION HILL RD	98053	<NA>	-122.0379	47.66734
## 9453	22340 NE 65TH PL	98053	<NA>	-122.0407	47.66467
## 9528	20338 NE 85TH ST	98053	<NA>	-122.0692	47.68341
## 9546	22465 NE 111TH LN	98053	<NA>	-122.0384	47.69732
## 9722	2850 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0927	47.63610
## 10125	20015 NE 85TH ST	98053	<NA>	-122.0729	47.67825
## 10318	8565 261ST AVE NE	98053	<NA>	-121.9914	47.67951
## 10371	22016 NE REDMOND FALL CITY RD	98053	<NA>	-122.0448	47.65178
## 10418	19841 NE REDMOND RD	98053	<NA>	-122.0716	47.69623
## 10478	5715 251ST CT NE	98053	<NA>	-122.0034	47.65602
## 10623	25621 NE 42ND PL	98053	<NA>	-121.9967	47.64604
## 10707	16410 NE 43RD CT	98052	REDMOND	-122.1208	47.64847
## 10723	15339 NE 106TH CT	98052	REDMOND	-122.1350	47.69446
## 10741	6228 224TH AVE NE	98053	<NA>	-122.0407	47.66293
## 10787	3230 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0972	47.63926
## 10844	15430 NE 51ST ST	98052	REDMOND	-122.1341	47.65443
## 10958	24143 NE 58TH PL	98053	<NA>	-122.0174	47.65712
## 10995	8513 255TH AVE NE	98053	<NA>	-121.9994	47.67765
## 11165	2410 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0878	47.63155
## 11289	3260 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0978	47.64009
## 11413	8071 208TH AVE NE	98053	<NA>	-122.0640	47.67567
## 11558	11004 132ND AVE NE	98052	REDMOND	-122.1642	47.69914
## 11586	2864 W LAKE SAMMAMISH PKWY NE	98052	REDMOND	-122.0929	47.63633
## 11728	21031 NE 122ND ST	98053	<NA>	-122.0590	47.70618
## 11758	22016 NE REDMOND FALL CITY RD	98053	<NA>	-122.0448	47.65178

## 11772	2808 W LAKE SAMMAMISH PKWY NE 98052	REDMOND	-122.0915	47.63512
## 11822	2420 W LAKE SAMMAMISH PKWY NE 98052	REDMOND	-122.0879	47.63179
## 11898	11207 248TH AVE NE 98053	<NA>	-122.0090	47.69878
## 11899	11207 248TH AVE NE 98053	<NA>	-122.0090	47.69878
## 11982	7350 259TH PL NE 98053	<NA>	-121.9927	47.66975
## 11992	17656 NE 116TH ST 98052	REDMOND	-122.1036	47.70440
## 12212	25205 NE 80TH ST 98053	<NA>	-122.0020	47.67283
## 12255	19610 NE 116TH ST 98053	<NA>	-122.0784	47.70355
## 12256	19610 NE 116TH ST 98053	<NA>	-122.0784	47.70355
## 12392	11220 196TH AVE NE 98053	<NA>	-122.0778	47.69907
## 12472	3020 W LAKE SAMMAMISH PKWY NE 98052	REDMOND	-122.0945	47.63747
## 12487	2656 W LAKE SAMMAMISH PKWY NE 98052	REDMOND	-122.0911	47.63458
## 12577	7702 196TH AVE NE 98053	<NA>	-122.0751	47.67266
## 12582	2942 W LAKE SAMMAMISH PKWY NE 98052	REDMOND	-122.0932	47.63718
## 12643	7917 219TH AVE NE 98053	<NA>	-122.0480	47.67454
## 12686	3851 E AMES LAKE DR NE 98053	<NA>	-121.9549	47.64393
## 12764	18025 NE 136TH ST 98052	<NA>	-122.0988	47.72113
##	building_grade square_feet_total_living bedrooms bath_full_count			
## 25	10 4920	4		4
## 108	9 4640	5		2
## 115	6 660	0		1
## 178	11 5800	5		4
## 239	9 3360	2		2
## 246	6 900	2		1
## 287	9 4710	4		2
## 295	11 5060	4		23
## 300	10 6880	5		1
## 341	11 4490	4		2
## 344	7 2700	3		1
## 359	11 5140	4		2
## 385	11 6310	4		2
## 396	12 5080	4		3
## 475	10 3320	4		1
## 482	12 6380	6		6
## 508	9 3700	4		3
## 528	9 5830	5		3
## 576	10 4320	4		3
## 661	12 8090	4		3
## 670	10 4710	4		2
## 679	11 8490	7		3
## 742	11 4577	4		2
## 784	10 5150	4		1
## 802	11 4380	4		3
## 811	11 5270	4		2
## 852	10 3630	4		2
## 853	9 3470	3		2
## 877	11 4640	4		3
## 916	12 7640	5		5
## 1009	9 4040	4		3
## 1099	11 4250	4		2
## 1119	11 6340	4		4
## 1142	11 5980	4		3
## 1155	8 4740	6		5
## 1305	8 2800	4		2

## 1345	10	3990	5	3
## 1368	10	4550	4	3
## 1380	8	3660	3	2
## 1442	4	340	0	0
## 1492	11	4610	4	4
## 1504	12	7780	5	3
## 1543	11	4310	4	3
## 1550	11	5380	4	3
## 1633	9	3130	5	2
## 1650	10	9360	4	3
## 1716	9	4610	4	2
## 1745	11	5360	4	3
## 1870	12	10630	5	4
## 1962	12	5300	4	3
## 1963	10	3090	3	1
## 1964	7	550	1	1
## 1976	9	3830	5	3
## 1977	9	3330	4	2
## 1978	9	3480	4	3
## 1979	9	3370	4	2
## 1980	9	3350	4	3
## 1981	9	4080	6	3
## 1982	9	3690	5	3
## 2022	7	900	1	1
## 2099	11	5020	4	3
## 2137	8	2820	3	1
## 2157	9	3130	4	2
## 2264	5	1540	3	1
## 2302	9	6340	10	2
## 2360	9	3200	3	1
## 2361	5	310	0	0
## 2469	8	2980	3	1
## 2684	10	3410	5	3
## 2685	9	3660	4	3
## 2686	10	4500	5	2
## 2687	9	3480	4	3
## 2688	10	3310	4	2
## 2689	10	4340	5	2
## 2690	10	4390	5	3
## 2699	9	5270	4	3
## 2708	9	4710	5	3
## 2709	10	5100	4	3
## 2710	9	4910	4	3
## 2717	12	5390	4	3
## 2742	10	3950	4	3
## 2852	13	11810	7	4
## 2934	9	4610	4	2
## 2937	9	4610	4	2
## 3097	12	5320	4	3
## 3102	11	4790	4	3
## 3110	6	1290	2	0
## 3111	7	1920	5	2
## 3168	7	1290	2	1
## 3169	7	1290	2	2

## 3170	7	1600	3	2
## 3171	7	1740	3	2
## 3172	7	1710	3	2
## 3173	7	1740	3	2
## 3174	7	1710	3	2
## 3175	7	1460	3	2
## 3176	7	1820	3	2
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## 3180	7	1840	3	2
## 3181	7	1600	3	2
## 3182	7	1290	2	1
## 3183	7	1290	2	1
## 3184	7	1290	2	1
## 3185	7	1290	2	1
## 3186	7	1290	2	1
## 3187	7	1290	2	1
## 3188	7	1290	2	2
## 3189	7	1460	3	2
## 3190	7	1710	3	2
## 3191	7	1460	3	2
## 3192	7	1740	3	2
## 3193	7	1710	3	2
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## 3195	7	1840	3	2
## 3196	7	1840	3	2
## 3197	7	1820	3	2
## 3198	7	1290	2	1
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## 3201	7	1290	2	1
## 3202	7	1290	2	1
## 3424	8	4290	4	3
## 3464	7	1460	3	2
## 3465	7	1600	3	2
## 3466	7	1840	3	2
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## 3470	7	1840	3	2
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## 3473	7	1710	3	2
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## 3478	7	1290	2	1
## 3479	7	1290	2	1
## 3480	7	1290	2	1
## 3481	7	1290	2	1
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## 3486	7	1600	3	2
## 3487	7	1840	3	2
## 3488	7	1710	3	2
## 3489	7	1740	3	2
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## 3491	7	1840	4	1
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## 3493	7	1840	3	2
## 3494	7	1460	3	2
## 3495	7	1290	2	1
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## 3523	8	3980	4	3
## 3837	9	4480	4	2
## 3918	11	6340	4	4
## 3919	11	5980	4	3
## 4055	11	4110	4	1
## 4056	8	900	1	1
## 4248	8	2610	3	2
## 4391	8	3140	2	2
## 4435	10	4330	4	2
## 4648	11	5790	3	2
## 4649	6	2410	3	1
## 4671	11	7810	5	5
## 4695	8	1430	2	1
## 4696	10	5330	4	3
## 4740	10	2900	3	2
## 4821	9	890	1	1
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## 4840	8	4740	6	5
## 4934	9	4210	4	2
## 5083	12	5830	4	4
## 5491	9	2960	4	2
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## 5497	9	2950	5	3
## 5549	10	4130	3	1
## 5935	10	5640	3	3
## 6055	9	3500	4	2
## 6230	10	3320	4	2
## 6231	10	3530	4	2
## 6232	10	3720	4	3
## 6233	10	3150	4	2
## 6234	10	3340	5	2
## 6235	10	3320	5	2
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## 6237	10	3320	5	2
## 6238	10	3060	4	2
## 6239	10	3530	4	2
## 6429	8	3290	4	2
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## 6431	8	2750	4	2
## 6432	8	3010	4	2
## 6433	8	3200	5	2
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## 6451	8	2520	4	3
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## 6512	12	5040	4	3
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## 6739	10	3850	3	4
## 6796	9	4000	4	3
## 6938	9	3970	4	2
## 6939	9	4610	4	3
## 6940	9	3360	4	2
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## 7447	8	2520	4	2
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## 7507	10	5150	4	1
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## 7791	11	5620	3	3
## 7871	12	6650	4	3
## 8154	9	3960	4	3
## 8232	10	5150	4	1
## 8262	12	13540	7	1
## 8320	10	5820	3	2
## 8377	12	8750	5	2
## 8457	10	3090	3	1
## 8458	7	550	1	1
## 8535	10	4420	5	2
## 8541	9	3130	5	2
## 8698	10	3380	4	2
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## 11558	5		730	2	1
## 11586	8		2770	4	1
## 11728	8		2280	2	1
## 11758	10		4930	4	0
## 11772	11		6600	4	2
## 11822	11		4240	3	2
## 11898	11		3690	4	2
## 11899	9		1230	1	1
## 11982	12		7640	5	5
## 11992	8		1670	3	1
## 12212	5		820	2	1
## 12255	5		1650	2	1
## 12256	8		3260	4	2
## 12392	7		2140	3	1
## 12472	8		3150	3	3
## 12487	9		5000	4	1
## 12577	10		5030	3	2
## 12582	12		7070	5	3
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## 12686	9		3720	3	2
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##	year_built	year_renovated	current_zoning	sq_ft_lot	
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## 108	1952	2004	11	19173	
## 115	1955	0	22	225640	
## 178	2008	0	21	63162	
## 239	2005	0	24	8752	
## 246	1918	0	11	14043	
## 287	2014	0	22	18498	
## 295	2016	0	22	89734	
## 300	2008	0	22	288367	
## 341	2008	0	11	55303	
## 344	2003	0	22	574992	
## 359	2008	0	22	212572	
## 385	2005	0	22	36362	
## 396	2005	0	22	180774	
## 475	1963	1989	11	10247	
## 482	2007	0	11	15021	
## 508	2010	0	22	305173	
## 528	2014	0	11	10454	
## 576	2008	0	22	223898	
## 661	2006	0	22	176418	
## 670	1992	0	11	15167	
## 679	2008	0	22	118483	
## 742	2011	0	22	98881	
## 784	2015	0	22	221720	
## 802	2008	0	22	221720	
## 811	2001	0	11	18045	
## 852	2007	0	22	243176	
## 853	1985	2001	22	182516	
## 877	2008	0	22	220413	
## 916	2007	0	22	144683	

## 1009	2005	0	22	35042
## 1099	2007	0	22	223027
## 1119	2007	0	22	45302
## 1142	2007	0	22	55756
## 1155	2007	0	22	544199
## 1305	1929	1993	11	10046
## 1345	2008	0	11	11842
## 1368	2010	0	22	204732
## 1380	2008	0	22	14820
## 1442	1954	0	22	29933
## 1492	2007	0	22	106722
## 1504	2008	0	17	266152
## 1543	2008	0	22	240855
## 1550	2007	0	22	106722
## 1633	2014	0	13	5930
## 1650	2009	0	21	45738
## 1716	2015	0	22	95989
## 1745	2006	0	19	46609
## 1870	2003	0	22	207781
## 1962	2006	0	22	179031
## 1963	1978	0	11	17715
## 1964	1979	0	11	17715
## 1976	2009	0	13	9480
## 1977	2009	0	13	7399
## 1978	2008	0	13	10309
## 1979	2008	0	13	7457
## 1980	2008	0	13	5788
## 1981	2008	0	13	8371
## 1982	2009	0	13	6731
## 2022	1940	0	11	18102
## 2099	2004	0	22	59677
## 2137	1972	0	11	17995
## 2157	1965	1997	11	13634
## 2264	1938	0	11	28183
## 2302	2009	0	11	17328
## 2360	1968	0	11	19556
## 2361	1964	1992	11	19556
## 2469	1980	0	11	19110
## 2684	2008	0	15	5794
## 2685	2010	0	15	5833
## 2686	2008	0	15	5169
## 2687	2010	0	15	5035
## 2688	2008	0	15	5650
## 2689	2008	0	15	5086
## 2690	2009	0	15	5476
## 2699	2015	0	23	1008414
## 2708	2015	0	22	108542
## 2709	2016	0	22	131301
## 2710	2015	0	22	146779
## 2717	2007	0	22	193842
## 2742	2013	0	22	78844
## 2852	2000	0	21	139392
## 2934	2015	0	22	95989
## 2937	2015	0	22	95989

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## 3102	2008	0	12	7701
## 3110	1961	0	11	18792
## 3111	1961	0	11	18792
## 3168	2008	0	24	2628
## 3169	2008	0	24	2628
## 3170	2009	0	24	6451
## 3171	2008	0	24	3833
## 3172	2008	0	24	4846
## 3173	2008	0	24	4862
## 3174	2008	0	24	4710
## 3175	2008	0	24	4350
## 3176	2009	0	24	4622
## 3177	2008	0	24	4697
## 3178	2008	0	24	5125
## 3179	2008	0	24	5481
## 3180	2008	0	24	5680
## 3181	2008	0	24	5277
## 3182	2008	0	24	2807
## 3183	2008	0	24	2482
## 3184	2008	0	24	2482
## 3185	2008	0	24	2828
## 3186	2008	0	24	2628
## 3187	2008	0	24	2628
## 3188	2008	0	24	3940
## 3189	2008	0	24	4093
## 3190	2008	0	24	4041
## 3191	2008	0	24	3989
## 3192	2008	0	24	3937
## 3193	2008	0	24	3885
## 3194	2008	0	24	3781
## 3195	2008	0	24	4910
## 3196	2008	0	24	5340
## 3197	2009	0	24	6952
## 3198	2008	0	24	3170
## 3199	2008	0	24	3140
## 3200	2008	0	24	2482
## 3201	2008	0	24	2482
## 3202	2008	0	24	2482
## 3424	2008	0	13	10922
## 3464	2008	0	24	3918
## 3465	2009	0	24	4036
## 3466	2008	0	24	4487
## 3467	2008	0	24	5453
## 3468	2008	0	24	4195
## 3469	2008	0	24	4179
## 3470	2008	0	24	4229
## 3471	2009	0	24	4279
## 3472	2008	0	24	3915
## 3473	2008	0	24	4424
## 3474	2008	0	24	4827
## 3475	2008	0	24	4948
## 3476	2008	0	24	4646
## 3477	2008	0	24	4607

## 3478	2008	0	24	2482
## 3479	2008	0	24	2482
## 3480	2008	0	24	2482
## 3481	2008	0	24	2482
## 3482	2008	0	24	2482
## 3483	2008	0	24	2483
## 3484	2008	0	24	2479
## 3485	2010	0	24	4408
## 3486	2008	0	24	4056
## 3487	2009	0	24	4424
## 3488	2008	0	24	3790
## 3489	2008	0	24	3476
## 3490	2008	0	24	4446
## 3491	2008	0	24	4140
## 3492	2009	0	24	4076
## 3493	2008	0	24	4698
## 3494	2008	0	24	4263
## 3495	2008	0	24	2483
## 3496	2008	0	24	2483
## 3497	2008	0	24	2483
## 3523	1955	1991	11	22545
## 3837	2014	0	23	104688
## 3918	2007	0	22	45302
## 3919	2007	0	22	55756
## 4055	1991	0	11	14388
## 4056	1937	1991	11	14388
## 4248	1988	0	22	228690
## 4391	2015	0	22	260488
## 4435	2011	0	13	11325
## 4648	1999	0	1	657816
## 4649	1935	0	2	1327090
## 4671	2011	0	22	277286
## 4695	2003	0	11	16105
## 4696	2005	0	11	16105
## 4740	1974	0	11	15188
## 4821	1999	0	13	7599
## 4834	1940	0	11	18102
## 4840	2007	0	22	544199
## 4934	1959	1989	11	15353
## 5083	1969	0	23	1127205
## 5491	2011	0	11	7743
## 5494	2011	0	11	6932
## 5495	2011	0	11	6396
## 5496	2011	0	11	7588
## 5497	2011	0	11	5815
## 5549	2013	0	22	104544
## 5935	2012	0	23	127591
## 6055	2013	0	22	223027
## 6230	2012	0	16	5000
## 6231	2014	0	16	5037
## 6232	2013	0	16	6237
## 6233	2013	0	16	5163
## 6234	2012	0	16	5013
## 6235	2012	0	16	5143

## 6236	2012	0	16	5100
## 6237	2013	0	16	5100
## 6238	2013	0	16	5103
## 6239	2014	0	16	4963
## 6429	2012	0	11	6712
## 6430	2010	0	11	4749
## 6431	2012	0	11	5816
## 6432	2012	0	11	8908
## 6433	2010	0	11	4584
## 6434	2012	0	11	4681
## 6435	2012	0	11	9901
## 6436	2012	0	11	13289
## 6437	2010	0	11	4368
## 6438	2011	0	11	4244
## 6439	2011	0	11	5778
## 6440	2012	0	11	6740
## 6441	2012	0	11	4451
## 6442	2012	0	11	5310
## 6443	2012	0	11	4647
## 6444	2012	0	11	4080
## 6445	2013	0	11	5032
## 6446	2013	0	11	4383
## 6447	2012	0	11	5326
## 6448	2011	0	11	5913
## 6449	2012	0	11	6254
## 6450	2012	0	11	5441
## 6451	2012	0	11	4442
## 6452	2013	0	11	4234
## 6453	2013	0	11	6168
## 6454	2012	0	11	4676
## 6455	2012	0	11	6695
## 6456	2012	0	11	7271
## 6457	2012	0	11	4596
## 6512	1990	0	11	17417
## 6527	2012	0	13	15681
## 6739	2013	0	22	62726
## 6796	2014	0	11	11780
## 6938	2013	0	11	6260
## 6939	2013	0	11	6260
## 6940	2013	0	11	6200
## 6941	2013	0	11	8108
## 6942	2013	0	11	6858
## 6943	2013	0	11	6475
## 6944	2013	0	11	6438
## 6945	2013	0	11	6260
## 6946	2013	0	11	6418
## 6947	2013	0	11	6594
## 6948	2013	0	11	7617
## 7039	1976	1998	19	151153
## 7147	2014	0	13	11718
## 7167	1975	2000	11	20119
## 7210	1978	2005	19	202387
## 7211	1982	0	19	155074
## 7446	2013	0	24	4644

## 7447	2013	0	24	4887
## 7448	2013	0	24	5078
## 7449	2013	0	24	4409
## 7450	2013	0	24	4400
## 7451	2013	0	24	4408
## 7452	2013	0	24	5140
## 7453	2012	0	24	5021
## 7454	2012	0	24	5099
## 7455	2012	0	24	5127
## 7456	2013	0	24	4850
## 7457	2013	0	24	4499
## 7458	2013	0	24	5232
## 7459	2013	0	24	4400
## 7460	2013	0	24	4400
## 7461	2013	0	24	4400
## 7462	2013	0	24	4403
## 7463	2013	0	24	4416
## 7507	2015	0	22	221720
## 7649	2014	0	14	8038
## 7650	2014	0	14	7796
## 7683	2008	0	11	29728
## 7791	2015	0	17	218439
## 7871	2015	0	22	319294
## 8154	2014	0	22	54014
## 8232	2015	0	22	221720
## 8262	1999	0	17	307752
## 8320	2016	0	18	436507
## 8377	1996	0	22	1631322
## 8457	1978	0	11	17715
## 8458	1979	0	11	17715
## 8535	2014	0	14	10000
## 8541	2014	0	13	5930
## 8698	2014	0	11	16190
## 8710	1996	0	11	14444
## 8763	1972	2007	11	10861
## 8887	2008	0	11	21010
## 8911	2003	0	6	83813
## 8946	2003	0	6	83813
## 9215	1954	0	1	316245
## 9293	2014	0	11	10454
## 9369	2015	0	22	279219
## 9420	2015	0	22	464350
## 9453	2015	0	22	95989
## 9528	2006	0	22	209088
## 9546	2014	0	23	104688
## 9722	1974	0	11	15188
## 10125	1998	0	22	167270
## 10318	1997	0	17	963702
## 10371	2016	0	22	111950
## 10418	2003	0	22	261795
## 10478	2015	0	22	93654
## 10623	2016	0	22	112384
## 10707	1987	0	11	9705
## 10723	2008	0	12	6451

## 10741	2015	0	22	30475
## 10787	1952	2004	11	19173
## 10844	1918	0	13	41789
## 10958	1997	0	19	117454
## 10995	2005	0	22	186525
## 11165	1995	0	11	14041
## 11289	1982	0	11	10532
## 11413	1948	0	22	192099
## 11558	1952	0	11	13260
## 11586	1938	0	11	13170
## 11728	1981	0	22	439956
## 11758	2016	0	22	111950
## 11772	2001	0	11	41217
## 11822	2005	0	11	15368
## 11898	1999	0	18	1166246
## 11899	1999	0	18	1166246
## 11982	2007	0	22	144683
## 11992	1964	0	11	425145
## 12212	1932	0	22	167838
## 12255	1980	0	22	227383
## 12256	1980	0	22	227383
## 12392	1987	0	22	188614
## 12472	1955	0	11	30894
## 12487	1962	1980	11	9600
## 12577	2005	0	22	227818
## 12582	1974	2003	11	29494
## 12643	2009	0	22	77418
## 12686	2006	0	22	19290
## 12764	2006	0	22	226512

b. xi. Investigate further by calculating the leverage, cooks distance, and

covariance ratios. Comment on all cases that are problematic.

```
# nrow(mydata) -- 12865

# Let's look at the leverage (hat value), cook's distance, and covariance ratio
# for cases with large.residual = TRUE
mydata[mydata$large.residual, c("cooks.distance", "leverage", "covariance.ratios")]
```

##	cooks.distance	leverage	covariance.ratios
## 25	0.0010257923	0.0011869228	0.9984429
## 108	0.0033044011	0.0042136523	1.0017920
## 115	0.0024117590	0.0018063670	0.9972755
## 178	0.0011465389	0.0012854582	0.9984365
## 239	0.0001090031	0.0001668056	0.9982225
## 246	0.0031564325	0.0017464582	0.9954241
## 287	0.0006893490	0.0007933254	0.9980305
## 295	0.2396277578	0.1188300220	1.1278061
## 300	0.0075267399	0.0043854071	0.9984257
## 341	0.0004487567	0.0006708699	0.9986685

## 344	0.0057351762	0.0089305205	1.0071144
## 359	0.0017397504	0.0014731451	0.9975264
## 385	0.0025307845	0.0015599477	0.9959382
## 396	0.0014805940	0.0014632775	0.9981591
## 475	0.0006986847	0.0009934765	0.9988611
## 482	0.0031738517	0.0039360295	1.0014278
## 508	0.0033052073	0.0053950353	1.0036382
## 528	0.0010875245	0.0016248737	0.9996238
## 576	0.0008319101	0.0013155085	0.9994538
## 661	0.0027968265	0.0028437415	0.9996540
## 670	0.0016660191	0.0005997724	0.9906016
## 679	0.0070450303	0.0036509393	0.9968763
## 742	0.0004484261	0.0007724219	0.9991065
## 784	0.0019524886	0.0023320386	0.9996953
## 802	0.0010311108	0.0015000453	0.9994297
## 811	0.0035835198	0.0009199438	0.9867072
## 852	0.0009102800	0.0015231355	0.9997943
## 853	0.0004455858	0.0007137136	0.9988811
## 877	0.0016204219	0.0014373881	0.9976943
## 916	0.0045527934	0.0030400158	0.9978991
## 1009	0.0002877654	0.0004295533	0.9984230
## 1099	0.0009422438	0.0015341461	0.9997424
## 1119	0.0020112581	0.0015651144	0.9972234
## 1142	0.0011810063	0.0010314525	0.9972200
## 1155	0.0149292319	0.0095001048	1.0041643
## 1305	0.0011346071	0.0016966438	0.9996980
## 1345	0.0002542520	0.0004161686	0.9986340
## 1368	0.0007075141	0.0011555917	0.9993700
## 1380	0.0003101548	0.0004637808	0.9984617
## 1442	0.0016275507	0.0015502173	0.9981018
## 1492	0.0010091685	0.0012623284	0.9987649
## 1504	0.0048822856	0.0030902960	0.9976377
## 1543	0.0010641713	0.0017134567	0.9998962
## 1550	0.0008218085	0.0008187770	0.9975437
## 1633	0.0001621965	0.0002170445	0.9979159
## 1650	0.0135138750	0.0062910557	0.9987168
## 1716	0.0009970945	0.0007971874	0.9965835
## 1745	0.0009432943	0.0007656760	0.9966233
## 1870	0.0074126080	0.0066798036	1.0030493
## 1962	0.0024740915	0.0014260536	0.9953770
## 1963	0.0006106903	0.0007478570	0.9981840
## 1964	0.0008404656	0.0004465255	0.9938367
## 1976	0.0013242658	0.0003997737	0.9883823
## 1977	0.0006853792	0.0001884214	0.9869458
## 1978	0.0012447634	0.0003547031	0.9875953
## 1979	0.0006774109	0.0001869084	0.9869939
## 1980	0.0012740900	0.0003552299	0.9873042
## 1981	0.0014775095	0.0004630181	0.9889040
## 1982	0.0012863954	0.0003793806	0.9880669
## 2022	0.0013424594	0.0009537856	0.9961454
## 2099	0.0005363480	0.0006876495	0.9982629
## 2137	0.0005589791	0.0003740237	0.9952344
## 2157	0.0005565648	0.0005217858	0.9970069
## 2264	0.0031592360	0.0015138722	0.9941316

## 2302	0.0017626301	0.0024168498	1.0001911
## 2360	0.0005459138	0.0005608369	0.9974005
## 2361	0.0024410539	0.0009953756	0.9922280
## 2469	0.0004473775	0.0003714912	0.9963336
## 2684	0.0015598153	0.0004735320	0.9885261
## 2685	0.0012316856	0.0003774574	0.9885464
## 2686	0.0014329427	0.0005141746	0.9904813
## 2687	0.0012141040	0.0003613572	0.9881655
## 2688	0.0011725802	0.0003451790	0.9880091
## 2689	0.0013120914	0.0004579103	0.9901284
## 2690	0.0013092871	0.0004699203	0.9904396
## 2699	0.0340776069	0.0254130949	1.0215325
## 2708	0.0013991091	0.0008301737	0.9949675
## 2709	0.0012136735	0.0008456224	0.9959317
## 2710	0.0017277029	0.0010943362	0.9956365
## 2717	0.0030168291	0.0015203003	0.9945262
## 2742	0.0005096100	0.0004644780	0.9968336
## 2852	0.0323551464	0.0084297003	0.9945063
## 2934	0.0009961450	0.0007971874	0.9965880
## 2937	0.0009613303	0.0007971874	0.9967538
## 3097	0.0026676183	0.0014175848	0.9948117
## 3102	0.0007357039	0.0006850037	0.9971425
## 3110	0.0008237369	0.0008555478	0.9977359
## 3111	0.0003968700	0.0005785581	0.9985112
## 3168	0.0039875726	0.0004821071	0.9699125
## 3169	0.0033779971	0.0004123887	0.9701391
## 3170	0.0029576959	0.0003738115	0.9711367
## 3171	0.0027292290	0.0003492111	0.9714720
## 3172	0.0027537800	0.0003513317	0.9713897
## 3173	0.0027317318	0.0003496169	0.9714795
## 3174	0.0027533928	0.0003512709	0.9713887
## 3175	0.0030548353	0.0003796849	0.9706469
## 3176	0.0027749988	0.0003587433	0.9717792
## 3177	0.0026791022	0.0003464883	0.9717788
## 3178	0.0026799098	0.0003466282	0.9717818
## 3179	0.0028631251	0.0003611646	0.9710675
## 3180	0.0026810696	0.0003468243	0.9717859
## 3181	0.0028623150	0.0003610450	0.9710660
## 3182	0.0039884401	0.0004822319	0.9699139
## 3183	0.0039868755	0.0004820066	0.9699114
## 3184	0.0039868755	0.0004820066	0.9699114
## 3185	0.0039885428	0.0004822466	0.9699141
## 3186	0.0039875726	0.0004821071	0.9699125
## 3187	0.0039875726	0.0004821071	0.9699125
## 3188	0.0033853118	0.0004134073	0.9701493
## 3189	0.0030536527	0.0003795151	0.9706449
## 3190	0.0027516005	0.0003509865	0.9713838
## 3191	0.0030531822	0.0003794474	0.9706441
## 3192	0.0027294619	0.0003492496	0.9714728
## 3193	0.0027512095	0.0003509237	0.9713827
## 3194	0.0027509545	0.0003508825	0.9713819
## 3195	0.0026794947	0.0003465567	0.9717803
## 3196	0.0026803440	0.0003467022	0.9717834
## 3197	0.0027812390	0.0003597503	0.9717962

## 3198	0.0039902426	0.0004824902	0.9699168
## 3199	0.0039900914	0.0004824686	0.9699165
## 3200	0.0039868755	0.0004820066	0.9699114
## 3201	0.0039868755	0.0004820066	0.9699114
## 3202	0.0039868755	0.0004820066	0.9699114
## 3424	0.0024495202	0.0040292686	1.0022766
## 3464	0.0029958323	0.0003794015	0.9712020
## 3465	0.0028920607	0.0003723198	0.9716726
## 3466	0.0026276922	0.0003464232	0.9723251
## 3467	0.0026294974	0.0003467421	0.9723320
## 3468	0.0029970621	0.0003795821	0.9712041
## 3469	0.0028044725	0.0003604398	0.9716123
## 3470	0.0026272739	0.0003463465	0.9723233
## 3471	0.0028929496	0.0003724558	0.9716744
## 3472	0.0026776797	0.0003492414	0.9720233
## 3473	0.0027005098	0.0003511464	0.9719381
## 3474	0.0029999885	0.0003800094	0.9712089
## 3475	0.0026801737	0.0003496534	0.9720308
## 3476	0.0037772871	0.0004924125	0.9721523
## 3477	0.0026793022	0.0003495111	0.9720283
## 3478	0.0039134029	0.0004820066	0.9704772
## 3479	0.0039134029	0.0004820066	0.9704772
## 3480	0.0039134029	0.0004820066	0.9704772
## 3481	0.0039134029	0.0004820066	0.9704772
## 3482	0.0039134029	0.0004820066	0.9704772
## 3483	0.0039134076	0.0004820073	0.9704772
## 3484	0.0033146796	0.0004122789	0.9707011
## 3485	0.0028051885	0.0003710987	0.9724468
## 3486	0.0028040584	0.0003603760	0.9716114
## 3487	0.0027138345	0.0003583941	0.9723855
## 3488	0.0026989172	0.0003508860	0.9719335
## 3489	0.0026767519	0.0003490836	0.9720201
## 3490	0.0027005680	0.0003511558	0.9719383
## 3491	0.0037826809	0.0004935759	0.9721801
## 3492	0.0028922054	0.0003723420	0.9716729
## 3493	0.0026280542	0.0003464886	0.9723266
## 3494	0.0029973689	0.0003796270	0.9712046
## 3495	0.0039134076	0.0004820073	0.9704772
## 3496	0.0039134076	0.0004820073	0.9704772
## 3497	0.0039134076	0.0004820073	0.9704772
## 3523	0.0026507975	0.0015849983	0.9957735
## 3837	0.0006838252	0.0007383763	0.9977580
## 3918	0.0012625486	0.0015651144	0.9990412
## 3919	0.0007652759	0.0010314525	0.9987521
## 4055	0.0014804592	0.0011854653	0.9969800
## 4056	0.0045005009	0.0013619354	0.9893770
## 4248	0.0007189969	0.0012086223	0.9994892
## 4391	0.0011440636	0.0019973401	1.0003650
## 4435	0.0003501356	0.0004602009	0.9981078
## 4648	0.1052345743	0.0102109306	0.9722188
## 4649	0.6906373336	0.0441092700	0.9884534
## 4671	0.0033942707	0.0042189274	1.0017201
## 4695	0.0015209988	0.0004222043	0.9873116
## 4696	0.0012415938	0.0008215703	0.9956192

## 4740	0.0003991572	0.0006797623	0.9989884
## 4821	0.0018888427	0.0015127881	0.9973094
## 4834	0.0011454391	0.0009537856	0.9969297
## 4840	0.0156021509	0.0095001048	1.0038956
## 4934	0.0009536065	0.0009580187	0.9977148
## 5083	0.0901980470	0.0309037562	1.0213738
## 5491	0.0001204817	0.0001827619	0.9982166
## 5494	0.0001184603	0.0001828554	0.9982600
## 5495	0.0001210330	0.0001833950	0.9982144
## 5496	0.0001205418	0.0001828299	0.9982163
## 5497	0.0002495714	0.0003894789	0.9984937
## 5549	0.0007267978	0.0011075487	0.9991545
## 5935	0.0013552866	0.0010634637	0.9967606
## 6055	0.0009548636	0.0013018514	0.9990555
## 6230	0.0016927822	0.0003674489	0.9834783
## 6231	0.0016557276	0.0003716195	0.9840513
## 6232	0.0018284484	0.0004261427	0.9847252
## 6233	0.0018688703	0.0003973082	0.9831440
## 6234	0.0016795075	0.0003655666	0.9835237
## 6235	0.0016921418	0.0003673263	0.9834790
## 6236	0.0016791110	0.0003654903	0.9835241
## 6237	0.0017228537	0.0003748257	0.9835249
## 6238	0.0019715318	0.0004140669	0.9829447
## 6239	0.0016560923	0.0003716927	0.9840510
## 6429	0.0051922484	0.0003579187	0.9468824
## 6430	0.0028656656	0.0001849280	0.9431183
## 6431	0.0034505351	0.0002283094	0.9445494
## 6432	0.0040677930	0.0002746190	0.9456881
## 6433	0.0044944781	0.0003068287	0.9463174
## 6434	0.0048150696	0.0003295367	0.9464729
## 6435	0.0087269903	0.0006216813	0.9488530
## 6436	0.0035389281	0.0002355183	0.9448734
## 6437	0.0028910251	0.0001879459	0.9435341
## 6438	0.0030345974	0.0001959032	0.9431503
## 6439	0.0044687097	0.0003045868	0.9462308
## 6440	0.0050615201	0.0003607200	0.9486138
## 6441	0.0072029810	0.0004873128	0.9460152
## 6442	0.0044465560	0.0003022794	0.9460878
## 6443	0.0062548245	0.0004833333	0.9527154
## 6444	0.0065752820	0.0005214943	0.9539694
## 6445	0.0066645597	0.0004997116	0.9512863
## 6446	0.0063454767	0.0004937736	0.9530550
## 6447	0.0071497350	0.0005789957	0.9549787
## 6448	0.0026806822	0.0002002402	0.9508024
## 6449	0.0059395065	0.0004822235	0.9549975
## 6450	0.0077410779	0.0006442085	0.9562607
## 6451	0.0065804276	0.0004930996	0.9512497
## 6452	0.0063582774	0.0004955632	0.9531320
## 6453	0.0076861191	0.0006311932	0.9556587
## 6454	0.0067255502	0.0005369915	0.9542924
## 6455	0.0062439332	0.0004799401	0.9524606
## 6456	0.0029151760	0.0002198457	0.9512864
## 6457	0.0047098882	0.0003741964	0.9539020
## 6512	0.0027687042	0.0013074985	0.9938082

## 6527	0.0011439540	0.0014878729	0.9991076
## 6739	0.0010574400	0.0010579402	0.9977993
## 6796	0.0002679505	0.0004499813	0.9987267
## 6938	0.0008439725	0.0003966984	0.9928578
## 6939	0.0012674214	0.0006937685	0.9942942
## 6940	0.0005499315	0.0002296020	0.9916774
## 6941	0.0005484610	0.0002296106	0.9917020
## 6942	0.0011257892	0.0005881841	0.9938582
## 6943	0.0021717622	0.0011535140	0.9945433
## 6944	0.0015445694	0.0008612923	0.9945893
## 6945	0.0008439725	0.0003966984	0.9928578
## 6946	0.0005520202	0.0002309316	0.9916967
## 6947	0.0005282048	0.0002164612	0.9914939
## 6948	0.0005193341	0.0002109403	0.9914059
## 7039	0.0008794222	0.0013133248	0.9993099
## 7147	0.0008071568	0.0008465522	0.9977625
## 7167	0.0040556743	0.0007040698	0.9794912
## 7210	0.0034382281	0.0013317753	0.9920764
## 7211	0.0049454782	0.0015271078	0.9897918
## 7446	0.0019642756	0.0002300190	0.9686643
## 7447	0.0019171957	0.0002216273	0.9682447
## 7448	0.0019661194	0.0002303729	0.9686836
## 7449	0.0019179172	0.0002216866	0.9682413
## 7450	0.0019172881	0.0002215034	0.9682251
## 7451	0.0019167196	0.0002213281	0.9682090
## 7452	0.0019168567	0.0002216008	0.9682465
## 7453	0.0018489967	0.0002157551	0.9685378
## 7454	0.0018541697	0.0002165786	0.9685709
## 7455	0.0018138062	0.0002095564	0.9682144
## 7456	0.0019172478	0.0002216315	0.9682445
## 7457	0.0019590073	0.0002291654	0.9686307
## 7458	0.0019156108	0.0002212414	0.9682150
## 7459	0.0019175169	0.0002199960	0.9679999
## 7460	0.0019172881	0.0002215034	0.9682251
## 7461	0.0019152975	0.0002203961	0.9680963
## 7462	0.0019160039	0.0002201497	0.9680481
## 7463	0.0019152776	0.0002203946	0.9680964
## 7507	0.0023158182	0.0024123317	0.9993094
## 7649	0.0006917948	0.0011398730	0.9993746
## 7650	0.0006918818	0.0011402519	0.9993754
## 7683	0.0103781570	0.0159671542	1.0143040
## 7791	0.0010461800	0.0015224008	0.9994529
## 7871	0.0046273592	0.0030157643	0.9977352
## 8154	0.0002961755	0.0004439867	0.9984482
## 8232	0.0024292806	0.0023320386	0.9989182
## 8262	0.0253133456	0.0181771335	1.0137703
## 8320	0.0064410706	0.0050596919	1.0007889
## 8377	0.3807445906	0.0644221375	1.0471077
## 8457	0.0005906999	0.0007478570	0.9982856
## 8458	0.0008222984	0.0004465255	0.9939909
## 8535	0.0003569741	0.0005152238	0.9984215
## 8541	0.0001631968	0.0002170445	0.9978983
## 8698	0.0002796865	0.0003719032	0.9980529
## 8710	0.0022473554	0.0007041959	0.9891443

```
## 8763    0.0005216706 0.0006359458    0.9980577
## 8887    0.0129194672 0.0027836422    0.9857741
## 8911    0.0013009607 0.0019755686    1.0000169
## 8946    0.0013009607 0.0019755686    1.0000169
## 9215    0.0024188561 0.0026122144    0.9996397
## 9293    0.0011160547 0.0016248737    0.9995570
## 9369    0.0013616896 0.0021917458    1.0003758
## 9420    0.0060749032 0.0073714500    1.0048367
## 9453    0.0005106974 0.0007971874    0.9989027
## 9528    0.0056161190 0.0022635893    0.9933926
## 9546    0.0006538835 0.0007383763    0.9979122
## 9722    0.0007277701 0.0006797623    0.9971501
## 10125   0.0052324495 0.0048478089    1.0013104
## 10318   0.0133567782 0.0224122117    1.0212129
## 10371   0.0021064306 0.0028288247    1.0005472
## 10418   0.0015299893 0.0023617801    1.0004458
## 10478   0.0004579563 0.0007820097    0.9990968
## 10623   0.0004269388 0.0006254808    0.9985711
## 10707   0.0001299349 0.0002014687    0.9982897
## 10723   0.0004312082 0.0006878271    0.9988453
## 10741   0.0002559881 0.0004202566    0.9986450
## 10787   0.0027445767 0.0014033335    0.9945172
## 10844   0.0016949044 0.0021391104    0.9996722
## 10958   0.0004796424 0.0004712638    0.9971420
## 10995   0.0030409613 0.0051210076    1.0034327
## 11165   0.0013206888 0.0014774990    0.9986218
## 11289   0.0006334948 0.0002389100    0.9907170
## 11413   0.0027417091 0.0012059380    0.9931160
## 11558   0.0042468752 0.0009956524    0.9853908
## 11586   0.0016574128 0.0011102733    0.9959786
## 11728   0.0068990834 0.0047846693    0.9998698
## 11758   0.0022656737 0.0028288247    1.0003331
## 11772   0.0120241217 0.0018168689    0.9774022
## 11822   0.0025447045 0.0006655231    0.9867224
## 11898   0.0327527392 0.0348098650    1.0330472
## 11899   0.0767419624 0.0375479305    1.0318106
## 11982   0.0086379247 0.0030400158    0.9928114
## 11992   0.0732808662 0.0046287603    0.9463640
## 12212   0.0020701946 0.0019228612    0.9983743
## 12255   0.0041240793 0.0022774808    0.9959444
## 12256   0.0011712201 0.0011821940    0.9979574
## 12392   0.0008750702 0.0010714059    0.9985077
## 12472   0.0042021297 0.0013192567    0.9897842
## 12487   0.0110948252 0.0046009420    0.9960107
## 12577   0.0039696302 0.0014696028    0.9917631
## 12582   0.0186385515 0.0024805710    0.9747313
## 12643   0.0160063267 0.0019543570    0.9716895
## 12686   0.0001473875 0.0002401013    0.9984466
## 12764   0.0054974616 0.0021949561    0.9932346
```

```
# check if any outlier cases have cook's distance > 1
sum(mydata[mydata$large.residual, c("cooks.distance")] > 1)
```

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

```

sum(mydata[, c("cooks.distance")] > 1)

## [1] 0
# None of the cases have cooks distance > 1, so none of the cases have undue
# influence on the model

# calculate average leverage using formula = (k+1/n)
avg_leverage <- (6+1)/12865
# twice the average leverage
times_2_leverage <- avg_leverage*2
# three times average leverage
times_3_leverage <- avg_leverage*3
# check if there are outlier cases with leverage > 3 times the average leverage
sum(mydata[mydata$large.residual, c("leverage")] >= times_3_leverage)

## [1] 0
# There are none. All the cases are within the range of three times the average.
sum(mydata[mydata$large.residual, c("leverage")] >= times_2_leverage)

## [1] 0
# There are none. All the cases are within the range of two times the average.

# check the minimum and maximum covariance ratios
# calculate the covariance minimum and maximum range
cov_ratio_min <- 1-(3*7/12865) # 1-[3(k+1)/n]
cov_ratio_max <- 1+(3*7/12865) # 1+[3(k+1)/n]
# checking covariance ratios
min(mydata[mydata$large.residual, c("covariance.ratios")])

## [1] 0.9431183
# 0.94 which is less than calculated minimum
max(mydata[mydata$large.residual, c("covariance.ratios")])

## [1] 1.127806
# 1.127 which is greater than calculated maximum
# check cook's distance for out of range covariance ratio cases
mydata$outside.covariance.ratios <- mydata$covariance.ratios > cov_ratio_max | mydata$covariance.ratios
# check cook's distance for out of range covariance ratios
sum(mydata[mydata$outside.covariance.ratios, c("cooks.distance")] > 1)

## [1] 0
# As there are no cases with cook's distance > 1 they are not alarming

```

b. xii. Perform the necessary calculations to assess the assumption of

independence and state if the condition is met or not.

```

# execute Durbin-Watson test to assess the assumption of independent error
library(car)

```



```
## Loading required package: carData
##
## Attaching package: 'car'
## The following object is masked from 'package:boot':
##
##     logit
## The following object is masked from 'package:dplyr':
##
##     recode
durbinWatsonTest(hs_data_multi_rgr_md1_4)

## lag Autocorrelation D-W Statistic p-value
## 1 0.7302629 0.5394659 0
## Alternative hypothesis: rho != 0
# Value is < 1 i.e. 0.539 and p-value is 0 which is < 0.05
# Result of the test conclude the residuals in the regression are positively
# correlated and thus fails the assumption of independence or errors.
library(lmtest)

## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##     as.Date, as.Date.numeric
dwtest(hs_data_multi_rgr_md1_4)

##
## Durbin-Watson test
##
## data: hs_data_multi_rgr_md1_4
## DW = 0.53947, p-value < 2.2e-16
## alternative hypothesis: true autocorrelation is greater than 0
# Alternative hypothesis is true and thus error terms or residuals are
# auto correlated.
```

b. xiii. Perform the necessary calculations to assess the assumption of no multicollinearity and state if the condition is met or not.

```
# calculating mean vif
mean(vif(hs_data_multi_rgr_md1_4))

## [1] 1.671316
# calculating vif
vif(hs_data_multi_rgr_md1_4)

##          sale_reason          building_grade square_feet_total_living
##          1.017011          2.362311          2.582797
##          bath_full_count          year_built          sq_ft_lot
```

```
##                1.576506                1.374726                1.114547
# calculating tolerance
1/vif(hs_data_multi_rgr_md1_4)

##                sale_reason                building_grade square_feet_total_living
##                0.9832737                0.4233143                0.3871772
##                bath_full_count                year_built                sq_ft_lot
##                0.6343139                0.7274178                0.8972257

# Observations
# 1. There is no vif greater than 10
# 2. Average vif is > 1, though not substantially and thus there may be some
# bias in the model (not significant). Value of vif is way below 5 and near 1.
# 3. None of the tolerance is below 0.1, no we do not have any serious problem
# with the model
# 4. None of the tolerance is below 0.2
# looking at above observation we can say that no multi-collinearity assumption
# is met.
```

b. xiv. Visually check the assumptions related to the residuals using the plot()

and hist() functions. Summarize what each graph is informing you of and if any

anomalies are present.

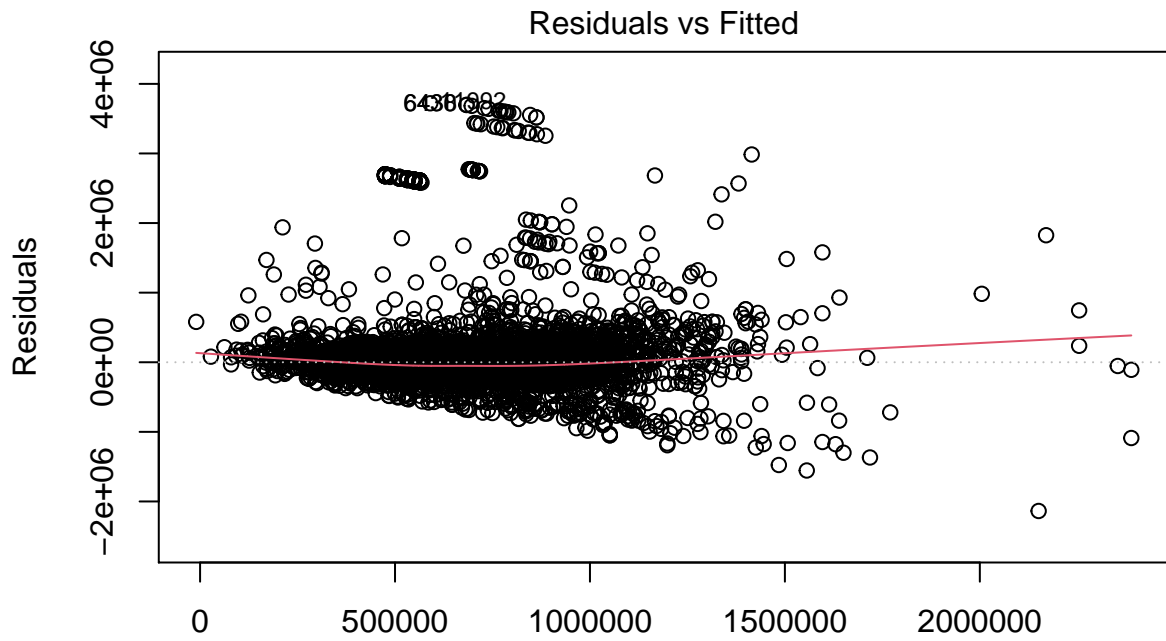
```
# Plotting standardized residual on y-axis and predicted values on x-axis.
# Plot is useful to determine if assumptions of random errors or homoscedasticity
# have been met.
# save fitted.values or predicted values as a variable in mydata data frame
mydata$fitted <- hs_data_multi_rgr_md1_4$fitted.values

library(ggplot2)
library(qqplotr)

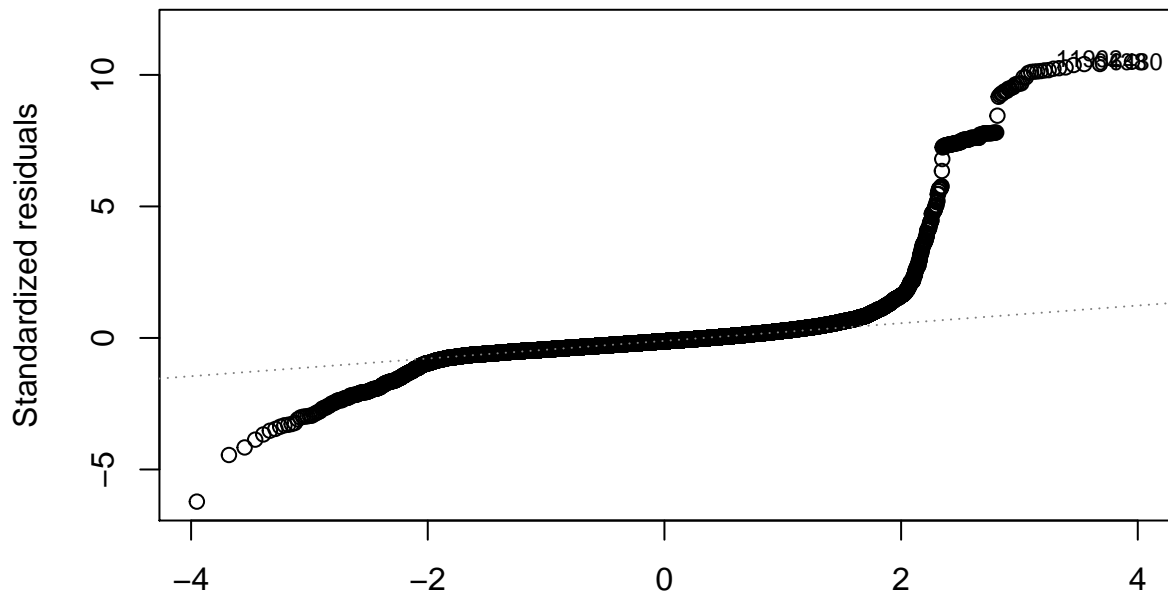
##
## Attaching package: 'qqplotr'

## The following objects are masked from 'package:ggplot2':
##
##      stat_qq_line, StatQqLine

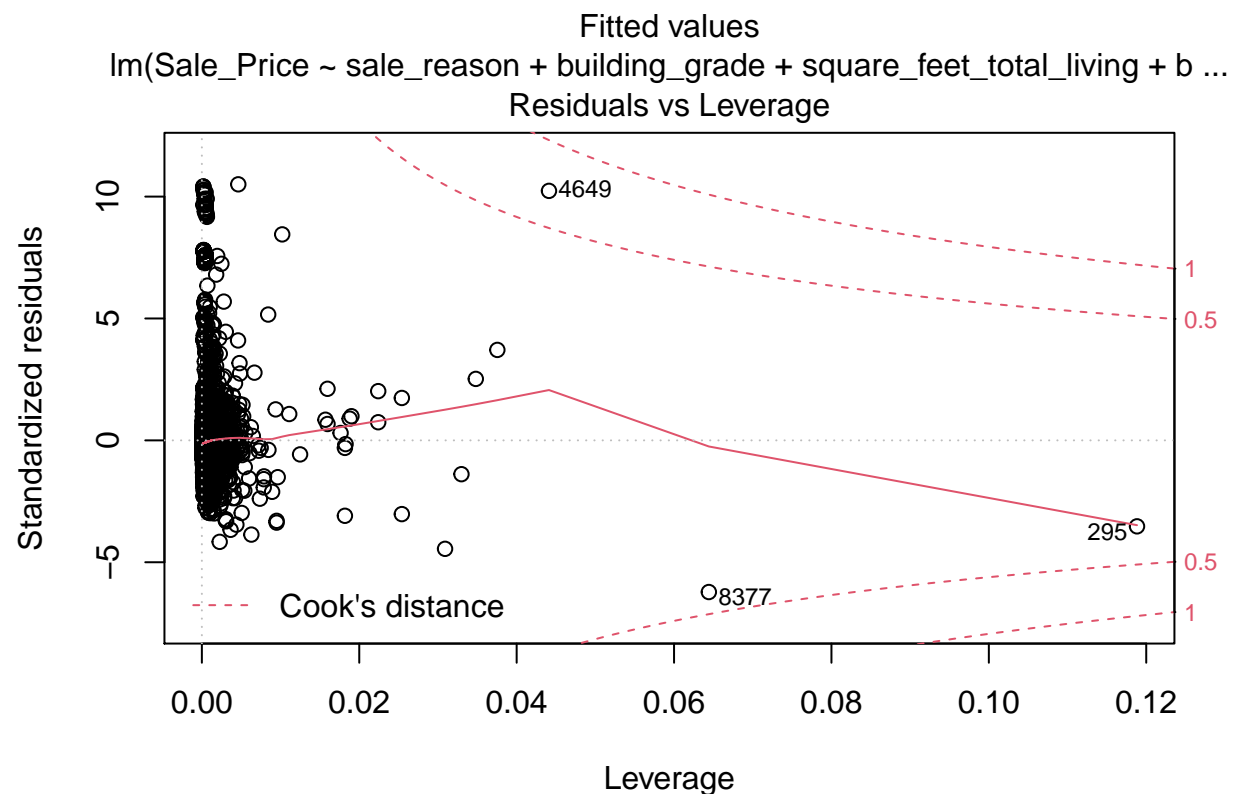
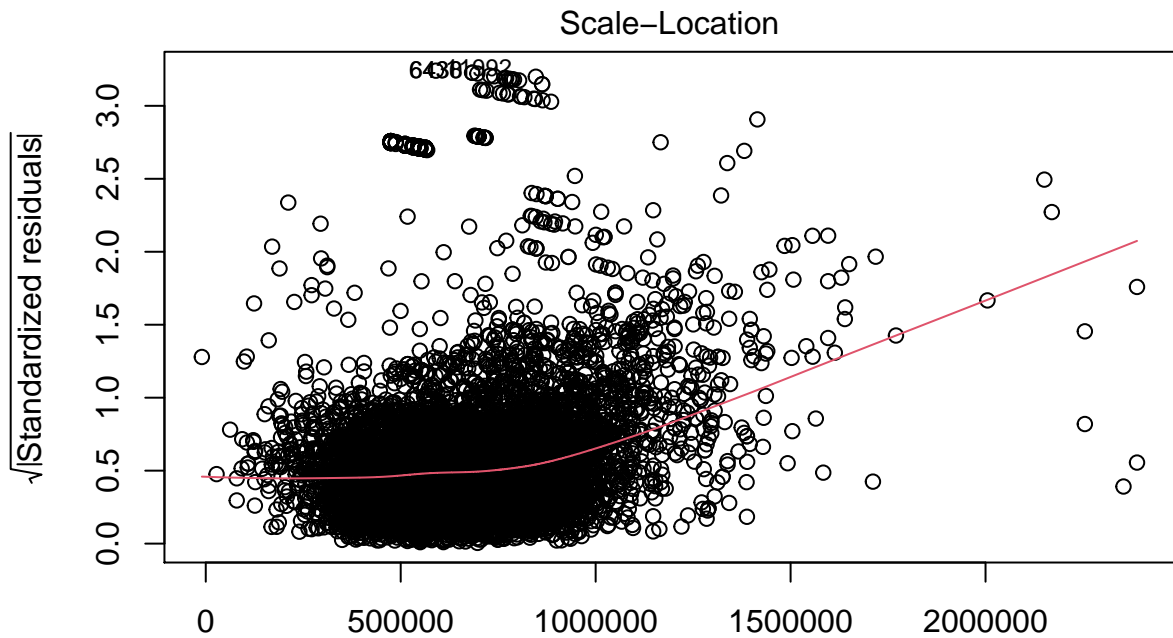
plot(hs_data_multi_rgr_md1_4)
```



Fitted values
 $\text{lm}(\text{Sale_Price} \sim \text{sale_reason} + \text{building_grade} + \text{square_feet_total_living} + \text{b} \dots)$
 Normal Q-Q



Theoretical Quantiles
 $\text{lm}(\text{Sale_Price} \sim \text{sale_reason} + \text{building_grade} + \text{square_feet_total_living} + \text{b} \dots)$



```
# Plot 1 - Scatterplot (Residual vs Fitted) - Assess linearity of data
# Ideally, the residual plot will show no fitted pattern. That is, the red
# line should be approximately horizontal at zero. The presence of a pattern may
# indicate a problem with some aspect of the linear model. In our example,
# there is no pattern in the residual plot. This suggests that we can assume
# linear relationship between the predictors and the outcome variables.
```

```

# Plot 2 = QQ plot of residuals - Access normality of residuals
# It can be used to visually check the normality assumption. The normal
# probability plot of residuals should approximately follow a straight line.
# In our example, half of the points fall exactly on the reference line which
# about half does not. We cannot assume normality of residuals in our case.

# Plot 3 - It's also called Spread-Location or scale-location plot.
# It is to evaluate Homogeneity of variance. This plot shows if residuals
# are spread equally along the ranges of predictors. It's good if you see a
# horizontal line with equally spread points. In our example, while majority
# of points are concentrated together and maintains the straight horizontal
# red line, few observations on right exist with higher residuals that are
# changes the red line from being horizontal to slant. I will still say overall
# data satisfies Homogeneity of variance because majority of points are randomly
# distributed and keeping the red line almost horizontal. If assumption is
# failed, a possible solution to reduce the heteroscedasticity problem is to
# use a log or square root transformation of the outcome variable (y).
# For e.g. model2 <- lm(log(sales) ~ youtube, data = marketing)

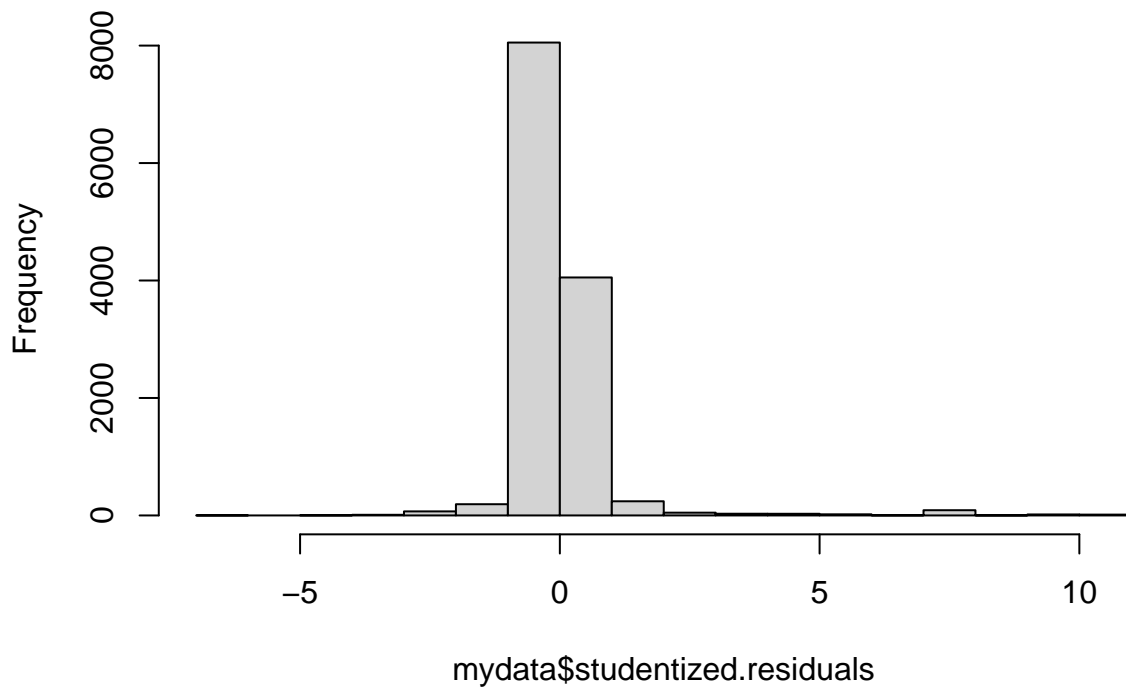
# Plot 4 - Residuals vs Leverage - outliers and high leverage points.
# Outliers - An outlier is a point that has an extreme outcome variable value.
# The presence of outliers may affect the interpretation of the model,
# because it increases the RSE. Outliers can be identified by examining the
# standardized residual (or studentized residual), which is the residual
# divided by its estimated standard error. Standardized residuals can be
# interpreted as the number of standard errors away from the regression line. Observations whose standard
# High leverage points - A data point has high leverage, if it has extreme
# predictor x values. This can be detected by examining the leverage statistic
# or the hat-value. A value of this statistic above  $2(k + 1)/n$  indicates an
# observation with high leverage, where k is the number of predictors and n
# is the number of observations.
# Influential values - An influential value is a value, which inclusion or
# exclusion can alter the results of the regression analysis. Such a value is
# associated with a large residual. Not all outliers (or extreme data points)
# are influential in linear regression analysis. Statisticians have developed
# a metric called Cook's distance to determine the influence of a value.
# This metric defines influence as a combination of leverage and residual size.
# Outliers and high leverage points can be identified by inspecting the
# Residuals vs Leverage plot.
# In our example, the plot highlights the top 3 most extreme points
# (#295, #4649 and #8377) with (-4,10,-5) approx standard residuals respectively.
# These are > 3 in absolute values and thus should be considered possible outliers.
# There are no points with higher leverage i.e. with leverage > than 2 times avg(leverage).
# The Residuals vs Leverage plot can help us to find influential observations if any.
# On this plot, outlying values are generally located at the upper right corner
# or at the lower right corner. Those spots are the places where data points
# can be influential against a regression line. We do not really have any
# points on the other side red dotted lines with Cook's distance >= 1.
# None on top right corner or bottom right corner. Thus, we do not see any
# outliers that are influential enough to change the regression line.

# plotting histograms of studentized.residuals and predicted values using hist()

```

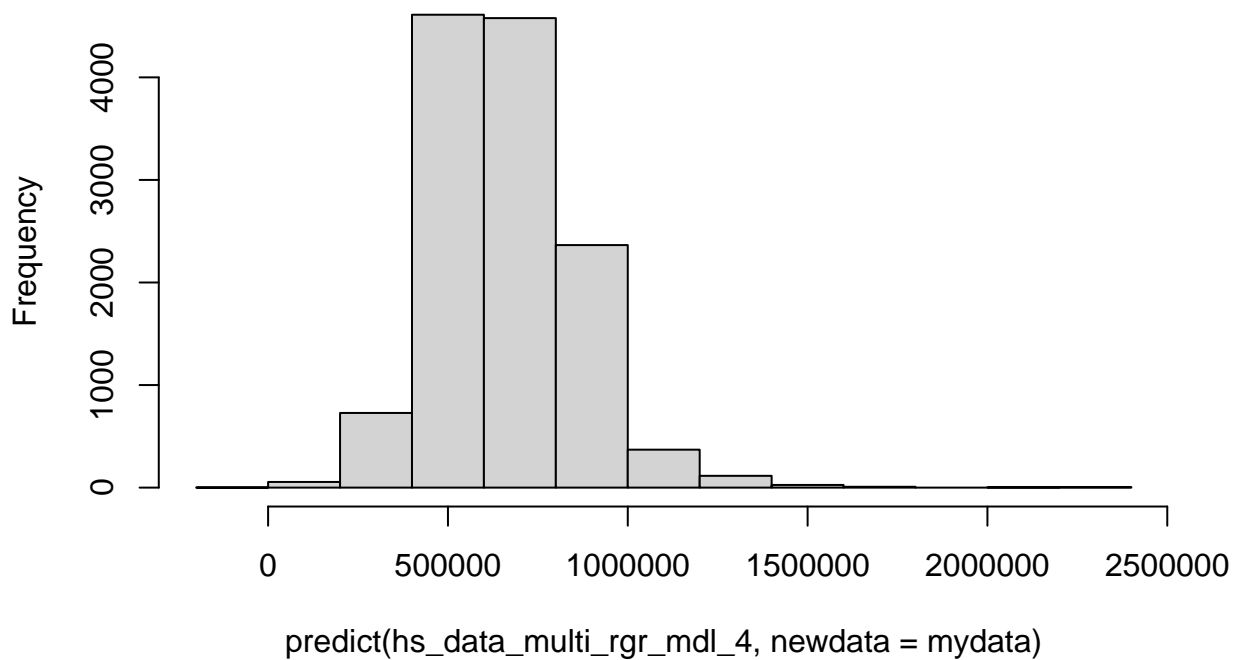
```
hist(mydata$studentized.residuals)
```

Histogram of mydata\$studentized.residuals



```
hist(predict(hs_data_multi_rgr_mdl_4, newdata = mydata))
```

Histogram of predict(hs_data_multi_rgr_mdl_4, newdata = mydata)



```
# as the histogram of residuals and predicted values have almost same shape,
# we can conclude that the residuals are normally distributed.
```

b. xv. Overall, is this regression model unbiased? If an unbiased regression model,

what does this tell us about the sample vs. the entire population model?

```
# 1. Utilizing vif values to gauge model bias
# When we check multi collinearity we check for vif score
vif(hs_data_multi_rgr_md1_4)
```

```
##           sale_reason           building_grade square_feet_total_living
##           1.017011           2.362311           2.582797
##           bath_full_count           year_built           sq_ft_lot
##           1.576506           1.374726           1.114547
```

```
# None of the vif scores are near 5 or greater and thus predictors does not
# have any significant multi collinearity. Multi collinearity problems consist of
# including, in the model, different variables that have a similar predictive
# relationship with the outcome.
```

```
# Average vif
mean(vif(hs_data_multi_rgr_md1_4))
```

```
## [1] 1.671316
```

```
# Average vif is >1 but nowhere close to 5 or greater. Model does not appear to have significant proof
```

```
# 2. We can also check bias by checking the impact of adding predictors on
# coefficient of first chosen predictor (in case of housing data - sq_ft_lot).
# If the coefficient of sq_ft_lot gets impacted much by addition of new predictors
# that means addition may cause bias.
```

```
# Coefficient of sq_ft_lot in simple regression model as base to compare
summary(hs_data_smpl_rgr_md1)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sq_ft_lot, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2016064  -194842   -63293    91565   3735109
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  6.418e+05  3.800e+03  168.90  <2e-16 ***
## sq_ft_lot    8.510e-01  6.217e-02  13.69  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
## Residual standard error: 401500 on 12863 degrees of freedom
## Multiple R-squared:  0.01435,    Adjusted R-squared:  0.01428
## F-statistic: 187.3 on 1 and 12863 DF,  p-value: < 2.2e-16
```

```
#           Estimate Std. Error t value Pr(>|t|)
#(Intercept) 6.418e+05  3.800e+03  168.90  <2e-16 ***
#sq_ft_lot   8.510e-01  6.217e-02   13.69  <2e-16 ***
```

```
# Trying to create models with one chosen variable at a time and observe change in coefficient of sq_ft.
hs_data_multi_rgr_mdl_5 <- lm(Sale_Price ~ sq_ft_lot+square_feet_total_living, data = mydata)
```

```
summary(hs_data_multi_rgr_mdl_5)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sq_ft_lot + square_feet_total_living,
##     data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1953643  -119763   -41295    44410   3770022
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    1.903e+05  8.771e+03  21.697  <2e-16 ***
## sq_ft_lot       1.006e-01  5.737e-02   1.754  0.0795 .
## square_feet_total_living 1.844e+02  3.300e+00  55.869  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 360200 on 12862 degrees of freedom
## Multiple R-squared:  0.2068, Adjusted R-squared:  0.2067
## F-statistic: 1677 on 2 and 12862 DF,  p-value: < 2.2e-16
```

```
#(Intercept)    1.903e+05  8.771e+03  21.697  <2e-16 ***
#sq_ft_lot       1.006e-01  5.737e-02   1.754  0.0795 .
#square_feet_total_living 1.844e+02  3.300e+00  55.869  <2e-16 ***
## difference caused in coefficient of sq_ft_lot (from simple regression) = 0.75
```

```
hs_data_multi_rgr_mdl_6 <- lm(Sale_Price ~ sq_ft_lot+building_grade, data = mydata)
```

```
summary(hs_data_multi_rgr_mdl_6)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sq_ft_lot + building_grade, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1943227  -135853   -48001    53756   3762342
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -5.147e+05  2.499e+04 -20.595  < 2e-16 ***
## sq_ft_lot     4.754e-01  5.805e-02   8.191 2.84e-16 ***
```



```
## building_grade 1.414e+05 3.025e+03 46.739 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 371200 on 12862 degrees of freedom
## Multiple R-squared:  0.1575, Adjusted R-squared:  0.1573
## F-statistic: 1202 on 2 and 12862 DF, p-value: < 2.2e-16

# (Intercept)    -5.147e+05  2.499e+04 -20.595 < 2e-16 ***
# sq_ft_lot      4.754e-01  5.805e-02  8.191 2.84e-16 ***
# building_grade 1.414e+05  3.025e+03 46.739 < 2e-16 ***
## difference caused in coefficient of sq_ft_lot (from simple regression) = 0.375

hs_data_multi_rgr_mdl_7 <- lm(Sale_Price ~ sq_ft_lot+bath_full_count, data = mydata)

summary(hs_data_multi_rgr_mdl_7)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sq_ft_lot + bath_full_count, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4128410 -161910  -49329    72518 3698383
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.312e+05  1.002e+04   33.05 <2e-16 ***
## sq_ft_lot      7.576e-01  5.973e-02   12.68 <2e-16 ***
## bath_full_count 1.739e+05  5.225e+03   33.28 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 385300 on 12862 degrees of freedom
## Multiple R-squared:  0.09249, Adjusted R-squared:  0.09235
## F-statistic: 655.4 on 2 and 12862 DF, p-value: < 2.2e-16

# (Intercept)    3.312e+05  1.002e+04   33.05 <2e-16 ***
# sq_ft_lot      7.576e-01  5.973e-02   12.68 <2e-16 ***
# bath_full_count 1.739e+05  5.225e+03   33.28 <2e-16 ***
# difference caused in coefficient of sq_ft_lot (from simple regression) = 0.1

hs_data_multi_rgr_mdl_8 <- lm(Sale_Price ~ sq_ft_lot+year_built, data = mydata)

summary(hs_data_multi_rgr_mdl_8)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sq_ft_lot + year_built, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2441124 -166193  -48805    74921 3634286
##
## Coefficients:
```

```
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.170e+07  3.991e+05 -29.32  <2e-16 ***
## sq_ft_lot   1.104e+00  6.054e-02  18.23  <2e-16 ***
## year_built  6.191e+03  2.002e+02  30.93  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 387400 on 12862 degrees of freedom
## Multiple R-squared:  0.08259,    Adjusted R-squared:  0.08245
## F-statistic: 579 on 2 and 12862 DF,  p-value: < 2.2e-16
```

```
# (Intercept) -1.170e+07  3.991e+05 -29.32  <2e-16 ***
# sq_ft_lot    1.104e+00  6.054e-02  18.23  <2e-16 ***
# year_built   6.191e+03  2.002e+02  30.93  <2e-16 ***
# difference caused in coefficient of sq_ft_lot (from simple regression) = -.25
```

```
hs_data_multi_rgr_mdl_9 <- lm(Sale_Price ~ sq_ft_lot+sale_reason, data = mydata)

summary(hs_data_multi_rgr_mdl_9)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sq_ft_lot + sale_reason, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2073083 -194249  -65029   86963  3725586
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  6.695e+05  4.265e+03  156.98  <2e-16 ***
## sq_ft_lot    8.802e-01  6.175e-02   14.25  <2e-16 ***
## sale_reason -1.827e+04  1.314e+03  -13.90  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 398500 on 12862 degrees of freedom
## Multiple R-squared:  0.02894,    Adjusted R-squared:  0.02879
## F-statistic: 191.7 on 2 and 12862 DF,  p-value: < 2.2e-16
```

```
# (Intercept)  6.695e+05  4.265e+03  156.98  <2e-16 ***
# sq_ft_lot     8.802e-01  6.175e-02   14.25  <2e-16 ***
# sale_reason -1.827e+04  1.314e+03  -13.90  <2e-16 ***
# difference caused in coefficient of sq_ft_lot (from simple regression) = -0.03
```

```
summary(hs_data_multi_rgr_mdl_4)
```

```
##
## Call:
## lm(formula = Sale_Price ~ sale_reason + building_grade + square_feet_total_living +
##      bath_full_count + year_built + sq_ft_lot, data = mydata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2136788 -119779  -44670   41079  3720748
```

```
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -4.477e+06  4.165e+05 -10.749 < 2e-16 ***
## sale_reason    -1.156e+04  1.180e+03  -9.792 < 2e-16 ***
## building_grade   3.106e+04  4.404e+03   7.053 1.84e-12 ***
## square_feet_total_living 1.374e+02  5.083e+00  27.038 < 2e-16 ***
## bath_full_count  1.288e+04  6.040e+03   2.132  0.033 *
## year_built      2.268e+03  2.132e+02  10.640 < 2e-16 ***
## sq_ft_lot       3.132e-01  5.805e-02   5.395 6.97e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 355100 on 12858 degrees of freedom
## Multiple R-squared:  0.2293, Adjusted R-squared:  0.2289
## F-statistic: 637.6 on 6 and 12858 DF,  p-value: < 2.2e-16

#(Intercept)          -4.477e+06  4.165e+05 -10.749 < 2e-16 ***
#sale_reason          -1.156e+04  1.180e+03  -9.792 < 2e-16 ***
#building_grade        3.106e+04  4.404e+03   7.053 1.84e-12 ***
#square_feet_total_living 1.374e+02  5.083e+00  27.038 < 2e-16 ***
#bath_full_count       1.288e+04  6.040e+03   2.132  0.033 *
#year_built            2.268e+03  2.132e+02  10.640 < 2e-16 ***
#sq_ft_lot             3.132e-01  5.805e-02   5.395 6.97e-08 ***
# difference caused in coefficient of sq_ft_lot (from simple regression) = 0.54

# As we can see that most difference caused to the coefficient of sq_ft_lot
# was when we added square_feet_total_living. Difference was 0.75. There may
# be some bias in the model because of this observation but it needs further
# analysis.

# 3. We can also observe correlation between variables of the model,
# specially checking correlation between sq_ft_lot and every other
# variable individually
cor(mydata[,c("sale_reason", "building_grade", "square_feet_total_living",
              "bath_full_count", "year_built", "sq_ft_lot")], method = "pearson")

##              sale_reason building_grade square_feet_total_living
## sale_reason          1.00000000    -0.07891229    -0.06521276
## building_grade       -0.07891229     1.00000000     0.74518052
## square_feet_total_living -0.06521276     0.74518052     1.00000000
## bath_full_count      -0.07312779     0.45941631     0.51778378
## year_built           -0.12023836     0.36192968     0.30642729
## sq_ft_lot            0.03400556     0.13843064     0.23410450
##              bath_full_count year_built  sq_ft_lot
## sale_reason       -0.07312779 -0.1202384  0.03400556
## building_grade     0.45941631  0.3619297  0.13843064
## square_feet_total_living 0.51778378  0.3064273  0.23410450
## bath_full_count    1.00000000  0.4525997  0.04699255
## year_built         0.45259966  1.0000000 -0.13491395
## sq_ft_lot          0.04699255 -0.1349139  1.00000000

# We see that there is some correlation positive correlation between
# sq_ft_lot and square_feet_total_living. We observed that addition of
# square_feet_total_living caused most change in the coefficient of sq_ft_lot as well.
```

```
# To conclude there may be little bias in the model because of correlation  
# between sq_ft_lot and square_feet_total_living. But as we see the results of  
# vif stats it is not significant enough.
```