

6.1 Assignment Housing Data

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```
##{r} ##setwd("/Users/Roni Kaakaty/Documents/Github/dsc520") ##
```

```
housing_df <- read.csv("data/week-7-cleaned-housing.csv")
```

```
library(ggplot2)
```

Explain why you chose to remove data points from your ‘clean’ dataset.

A. Not applicable since we received “cleaned” data set already. However, I did update the half bathroom and 3 quarter bathroom columns. I wanted a total_bath variable, so in order to weigh the bathrooms equally, every 1 half bath became 0.5 and every 1 three quarter bath became 0.75. This will enable me to provide a more accurate representation for my prediction model.

Create two variables; one that will contain variables Sale Price and Square Foot of Lot and one that will contain Sale Price and several additional predictors of your choice. Explain the basis of your additional predictor selections.

B. I’ve included square feet total living, bedrooms and total baths as additional predictor selections as there is an assumption that those items have a strong correlation with the sales price.

##Create two variables

```
sqfoot_lm <- lm(Sale.Price ~ sq_ft_lot, data = housing_df)
```

```
saleprice_lm <- lm(Sale.Price ~ sq_ft_lot + square_feet_total_living + bedrooms + total_bath, data = housing_df)
```

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?

C.

Sale Price/Sq_ft_lot:

R2: .01435 Adjusted R2: .01428

Sale Price/Sq_ft_lot/Bedroom/Total Bath/Sq_ft_living:

R2: .2098 Adjusted R2: .2096

These results tell me that a linear model might not be the best fit since the numbers are very far from 1. Yes, including additional parameters is helpful because it accounted for variance went from 1% in the original model to 21% in the model with the additional parameters. This means that the parameters helped account for an additional 20%.

```
summary(sqfoot_lm)
```

```
##
## Call:
## lm(formula = Sale.Price ~ sq_ft_lot, data = housing_df)
##
## 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
```

```
summary(saleprice_lm)
```

```
##
## Call:
## lm(formula = Sale.Price ~ sq_ft_lot + square_feet_total_living +
##      bedrooms + total_bath, data = housing_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1975012  -117703   -40353    44570   3787149
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.158e+05  1.451e+04  14.869  < 2e-16 ***
## sq_ft_lot         9.441e-02  5.800e-02   1.628    0.104
## square_feet_total_living  1.828e+02  5.268e+00  34.698  < 2e-16 ***
## bedrooms        -2.734e+04  4.506e+03  -6.068  1.33e-09 ***
## total_bath        2.979e+04  6.969e+03   4.275  1.92e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 359500 on 12860 degrees of freedom
## Multiple R-squared:  0.2098, Adjusted R-squared:  0.2096
## F-statistic: 853.6 on 4 and 12860 DF, p-value: < 2.2e-16
```

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?

D. I've populated the standardized betas below. They indicate that for every 1 standard deviation in the predictor, the coefficient in question will increase by their standard beta coefficient, assuming the other variables remain constant.

```
###{r} ##install.packages("QuantPsyc") ##
```

```
library(QuantPsyc)
```

```
## Warning: package 'QuantPsyc' was built under R version 4.0.2
```

```
## Loading required package: boot
```

```
## Loading required package: MASS
```

```
##
```

```
## Attaching package: 'QuantPsyc'
```

```
## The following object is masked from 'package:base':
```

```
##
```

```
##      norm
```

```
lm.beta(saleprice_lm)
```

```
##              sq_ft_lot square_feet_total_living      bedrooms
##              0.01329249          0.44745161          -0.05923540
##              total_bath
##              0.05121769
```

Calculate the confidence intervals for the parameters in your model and explain what the results indicate.

```
confint(saleprice_lm)
```

```
##              2.5 %      97.5 %
## (Intercept)    1.873207e+05  2.442088e+05
## sq_ft_lot      -1.927802e-02  2.081036e-01
## square_feet_total_living  1.724756e+02  1.931291e+02
## bedrooms      -3.617257e+04 -1.850824e+04
## total_bath      1.613289e+04  4.345214e+04
```

E. The results indicate that the true value of the unknown quantities should fall within the ranges listed with 95% confidence. Since the confidence interval for the square_feet_total_living has the tightest confidence interval, it indicates that the estimates for the model should be representative of the true population.

Assess the improvement of the new model compared to the original model by testing whether this change is significant by performing an analysis of variance.

F. The analysis of variance provided a F value of 1060.2 which confirms that the new model improved the old model significantly.

```
anova(sqfoot_lm, saleprice_lm)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Model 1: Sale.Price ~ sq_ft_lot
```

```
## Model 2: Sale.Price ~ sq_ft_lot + square_feet_total_living + bedrooms +
```

```
##      total_bath
```

```
##      Res.Df      RSS Df Sum of Sq      F      Pr(>F)
## 1   12863 2.0734e+15
## 2   12860 1.6622e+15  3 4.1113e+14 1060.2 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

G. Perform casewise diagnostics to identify outliers and/or influential cases, storing each function's output in a dataframe assigned to a unique variable frame.

```
housing_df$residuals <- resid(saleprice_lm)
```

H. Calculate the standardized residuals using the appropriate command, specifying those that are ± 2 , storing the results of large residuals in a variable you create.

```
housing_df$standardized.residuals <- rstandard(saleprice_lm)
housing_df$studentized.residuals <- rstudent(saleprice_lm)
housing_df$cooks.distance <- cooks.distance(saleprice_lm)
housing_df$dfbeta <- dfbeta(saleprice_lm)
housing_df$dffit <- dffits(saleprice_lm)
housing_df$leverage <- hatvalues(saleprice_lm)
housing_df$covariance.ratios <- covratio(saleprice_lm)
```

```
write.table(housing_df, "Housing with Diagnostics.csv", sep = "\t", row.names = FALSE)
```

```
housing_df$large.residual <- housing_df$standardized.residuals > 2 | housing_df$standardized.residuals
```

I. Use the appropriate function to show the sum of the large residuals.

```
sum(housing_df$large.residual)
```

```
## [1] 322
```

J. Which specific variables have large residuals (only cases that evaluate as TRUE?)

```
housing_df[housing_df$large.residual, c("bedrooms", "total_bath", "sq_ft_lot", "square_feet_total_living")]
```

```
##      bedrooms total_bath sq_ft_lot square_feet_total_living
## 6           4      3.25      7280             4160
## 25          4      4.50     112650             4920
## 115         0      1.00     225640              660
## 178         5      4.50      63162             5800
## 239         2      2.50       8752             3360
## 246         2      1.00      14043              900
## 287         4      4.00      18498             4710
## 295         4     23.50      89734             5060
## 300         5      4.50     288367             6880
## 341         4      3.25      55303             4490
## 359         4      5.50     212572             5140
## 385         4      3.75      36362             6310
## 396         4      4.00     180774             5080
```

## 475	4	3.00	10247	3320
## 482	6	6.50	15021	6380
## 508	4	3.75	305173	3700
## 528	5	5.00	10454	5830
## 661	4	4.25	176418	8090
## 670	4	3.25	15167	4710
## 679	7	7.75	118483	8490
## 784	4	3.00	221720	5150
## 811	4	4.25	18045	5270
## 877	4	3.50	220413	4640
## 916	5	5.50	144683	7640
## 1009	4	3.00	35042	4040
## 1119	4	5.25	45302	6340
## 1142	4	4.25	55756	5980
## 1155	6	5.75	544199	4740
## 1207	4	4.75	37388	8610
## 1380	3	2.00	14820	3660
## 1442	0	0.75	29933	340
## 1492	4	4.50	106722	4610
## 1504	5	4.75	266152	7780
## 1550	4	4.00	106722	5380
## 1633	5	2.75	5930	3130
## 1650	4	4.25	45738	9360
## 1716	4	3.50	95989	4610
## 1745	4	4.25	46609	5360
## 1870	5	5.25	207781	10630
## 1962	4	3.50	179031	5300
## 1963	3	3.00	17715	3090
## 1964	1	1.00	17715	550
## 1976	5	3.75	9480	3830
## 1977	4	2.50	7399	3330
## 1978	4	3.00	10309	3480
## 1979	4	2.50	7457	3370
## 1980	4	3.50	5788	3350
## 1981	6	3.50	8371	4080
## 1982	5	3.50	6731	3690
## 2022	1	1.75	18102	900
## 2099	4	4.25	59677	5020
## 2137	3	2.25	17995	2820
## 2157	4	3.25	13634	3130
## 2257	3	2.50	23522	2940
## 2264	3	1.75	28183	1540
## 2302	10	8.25	17328	6340
## 2360	3	2.50	19556	3200
## 2361	0	0.00	19556	310
## 2469	3	2.50	19110	2980
## 2604	3	3.75	41583	3060
## 2684	5	3.75	5794	3410
## 2685	4	3.50	5833	3660
## 2686	5	4.00	5169	4500
## 2687	4	3.50	5035	3480
## 2688	4	3.25	5650	3310
## 2689	5	3.25	5086	4340
## 2690	5	4.25	5476	4390

## 2699	4	3.75	1008414	5270
## 2708	5	4.25	108542	4710
## 2709	4	3.50	131301	5100
## 2710	4	3.50	146779	4910
## 2717	4	3.50	193842	5390
## 2742	4	4.25	78844	3950
## 2852	7	7.50	139392	11810
## 2934	4	3.50	95989	4610
## 2937	4	3.50	95989	4610
## 3097	4	3.50	178160	5320
## 3102	4	3.50	7701	4790
## 3110	2	1.25	18792	1290
## 3111	5	2.00	18792	1920
## 3168	2	1.00	2628	1290
## 3169	2	2.00	2628	1290
## 3170	3	2.50	6451	1600
## 3171	3	2.50	3833	1740
## 3172	3	2.50	4846	1710
## 3173	3	2.00	4862	1740
## 3174	3	2.50	4710	1710
## 3175	3	2.00	4350	1460
## 3176	3	2.50	4622	1820
## 3177	3	2.50	4697	1840
## 3178	3	2.00	5125	1840
## 3179	3	2.50	5481	1600
## 3180	3	2.50	5680	1840
## 3181	3	2.50	5277	1600
## 3182	2	1.00	2807	1290
## 3183	2	1.00	2482	1290
## 3184	2	1.00	2482	1290
## 3185	2	1.00	2828	1290
## 3186	2	1.00	2628	1290
## 3187	2	1.00	2628	1290
## 3188	2	2.00	3940	1290
## 3189	3	2.00	4093	1460
## 3190	3	2.50	4041	1710
## 3191	3	2.00	3989	1460
## 3192	3	2.50	3937	1740
## 3193	3	2.50	3885	1710
## 3194	3	2.50	3781	1710
## 3195	3	2.50	4910	1840
## 3196	3	2.50	5340	1840
## 3197	3	2.50	6952	1820
## 3198	2	1.00	3170	1290
## 3199	2	1.00	3140	1290
## 3200	2	1.00	2482	1290
## 3201	2	1.00	2482	1290
## 3202	2	1.00	2482	1290
## 3424	4	3.50	10922	4290
## 3464	3	2.00	3918	1460
## 3465	3	2.50	4036	1600
## 3466	3	2.50	4487	1840
## 3467	3	2.50	5453	1840
## 3468	3	2.00	4195	1460

## 3469	3	2.50	4179	1600
## 3470	3	2.50	4229	1840
## 3471	3	2.50	4279	1600
## 3472	3	2.50	3915	1740
## 3473	3	2.50	4424	1710
## 3474	3	2.00	4827	1460
## 3475	3	2.50	4948	1740
## 3476	3	1.50	4646	1830
## 3477	3	2.50	4607	1740
## 3478	2	1.00	2482	1290
## 3479	2	1.00	2482	1290
## 3480	2	1.00	2482	1290
## 3481	2	1.00	2482	1290
## 3482	2	1.00	2482	1290
## 3483	2	1.00	2483	1290
## 3484	2	2.00	2479	1290
## 3485	3	2.50	4408	1840
## 3486	3	2.50	4056	1600
## 3487	3	2.50	4424	1840
## 3488	3	2.50	3790	1710
## 3489	3	2.50	3476	1740
## 3490	3	2.00	4446	1710
## 3491	4	2.25	4140	1840
## 3492	3	2.00	4076	1600
## 3493	3	2.50	4698	1840
## 3494	3	2.00	4263	1460
## 3495	2	1.00	2483	1290
## 3496	2	1.00	2483	1290
## 3497	2	1.00	2483	1290
## 3523	4	4.25	22545	3980
## 3810	5	2.75	16418	3430
## 3837	4	2.50	104688	4480
## 3918	4	5.25	45302	6340
## 3919	4	4.25	55756	5980
## 4055	4	4.50	14388	4110
## 4056	1	1.00	14388	900
## 4285	5	3.00	8561	4180
## 4435	4	4.00	11325	4330
## 4571	3	1.50	105660	3430
## 4648	3	3.25	657816	5790
## 4649	3	1.75	1327090	2410
## 4671	5	7.00	277286	7810
## 4695	2	1.75	16105	1430
## 4696	4	3.50	16105	5330
## 4821	1	1.00	7599	890
## 4834	1	1.75	18102	900
## 4840	6	5.75	544199	4740
## 4934	4	2.75	15353	4210
## 5083	4	4.75	1127205	5830
## 5491	4	2.50	7743	2960
## 5494	4	2.50	6932	3010
## 5495	4	2.75	6396	2960
## 5496	4	2.50	7588	2960
## 5497	5	3.00	5815	2950

## 5498	4	2.50	5062	3480
## 5549	3	2.75	104544	4130
## 5935	3	3.50	127591	5640
## 6055	4	2.75	223027	3500
## 6230	4	3.25	5000	3320
## 6231	4	3.25	5037	3530
## 6232	4	3.50	6237	3720
## 6233	4	2.75	5163	3150
## 6234	5	3.50	5013	3340
## 6235	5	2.75	5143	3320
## 6236	5	3.50	5100	3340
## 6237	5	2.75	5100	3320
## 6238	4	2.75	5103	3060
## 6239	4	3.25	4963	3530
## 6429	4	2.50	6712	3290
## 6430	4	2.50	4749	2450
## 6431	4	2.50	5816	2750
## 6432	4	2.75	8908	3010
## 6433	5	2.50	4584	3200
## 6434	5	2.50	4681	3200
## 6435	5	3.75	9901	3620
## 6436	4	2.50	13289	2810
## 6437	4	2.50	4368	2550
## 6438	3	2.50	4244	2440
## 6439	4	2.50	5778	3160
## 6440	4	3.75	6740	3400
## 6441	4	3.75	4451	2960
## 6442	4	2.75	5310	3110
## 6443	5	3.50	4647	2900
## 6444	5	3.50	4080	3220
## 6445	4	3.50	5032	2510
## 6446	4	3.50	4383	2970
## 6447	5	3.50	5326	3470
## 6448	4	2.50	5913	2580
## 6449	5	3.25	6254	3590
## 6450	5	3.25	5441	3890
## 6451	4	3.50	4442	2520
## 6452	4	3.50	4234	2990
## 6453	5	3.00	6168	3620
## 6454	5	3.50	4676	3300
## 6455	5	3.00	6695	2830
## 6456	4	2.50	7271	2680
## 6457	5	3.25	4596	3330
## 6512	4	4.25	17417	5040
## 6527	5	3.75	15681	5050
## 6634	5	2.50	5898	2670
## 6739	3	4.50	62726	3850
## 6821	5	2.50	16000	4200
## 6931	4	2.50	7761	4000
## 6938	4	2.75	6260	3970
## 6939	4	3.75	6260	4610
## 6940	4	2.50	6200	3360
## 6941	4	2.50	8108	3370
## 6942	5	3.75	6858	4390

## 6943	5	4.00	6475	4380
## 6944	5	3.50	6438	4790
## 6945	4	3.50	6260	3970
## 6946	4	2.50	6418	3370
## 6947	4	2.50	6594	3260
## 6948	4	2.50	7617	3210
## 7039	4	3.50	151153	5320
## 7147	5	3.50	11718	5370
## 7167	4	2.75	20119	4460
## 7210	4	3.00	202387	4460
## 7211	5	3.00	155074	3640
## 7389	3	4.25	79863	7040
## 7446	4	2.50	4644	2650
## 7447	4	2.50	4887	2520
## 7448	4	2.50	5078	2655
## 7449	3	2.50	4409	2520
## 7450	3	2.50	4400	2515
## 7451	3	2.50	4408	2510
## 7452	4	2.50	5140	2520
## 7453	4	2.50	5021	2630
## 7454	4	2.50	5099	2640
## 7455	4	2.50	5127	2530
## 7456	4	2.50	4850	2520
## 7457	4	2.50	4499	2640
## 7458	3	2.50	5232	2510
## 7459	4	2.50	4400	2445
## 7460	3	2.50	4400	2515
## 7461	4	2.50	4400	2475
## 7462	4	2.50	4403	2460
## 7463	4	2.50	4416	2475
## 7507	4	3.00	221720	5150
## 7649	4	4.50	8038	4350
## 7650	4	4.50	7796	4350
## 7791	3	4.00	218439	5620
## 7871	4	4.25	319294	6650
## 8119	5	2.50	16000	4200
## 8154	4	3.50	54014	3960
## 8232	4	3.00	221720	5150
## 8262	7	8.00	307752	13540
## 8320	3	3.25	436507	5820
## 8377	5	5.25	1631322	8750
## 8457	3	3.00	17715	3090
## 8458	1	1.00	17715	550
## 8535	5	3.25	10000	4420
## 8541	5	2.75	5930	3130
## 8698	4	3.25	16190	3380
## 8710	5	4.25	14444	4950
## 8763	5	2.50	10861	3460
## 8887	4	5.50	21010	6360
## 8911	5	5.00	83813	6280
## 8946	5	5.00	83813	6280
## 9215	4	1.00	316245	1650
## 9293	5	5.00	10454	5830
## 9420	4	0.50	464350	4800

## 9453	4	3.50	95989	4610
## 9528	4	4.00	209088	7110
## 9546	4	2.50	104688	4480
## 9722	3	2.75	15188	2900
## 10125	4	3.50	167270	6680
## 10318	4	3.75	963702	5150
## 10371	4	2.25	111950	4930
## 10418	3	3.00	261795	4770
## 10623	4	3.50	112384	4570
## 10707	4	2.50	9705	2630
## 10723	4	3.50	6451	4790
## 10787	5	4.25	19173	4640
## 10958	5	3.25	117454	4010
## 10995	6	6.00	186525	9070
## 11165	4	4.50	14041	4140
## 11289	4	3.25	10532	3310
## 11413	3	2.25	192099	2610
## 11558	2	1.00	13260	730
## 11586	4	2.50	13170	2770
## 11728	2	1.75	439956	2280
## 11758	4	2.25	111950	4930
## 11772	4	3.25	41217	6600
## 11822	3	3.25	15368	4240
## 11898	4	4.00	1166246	3690
## 11899	1	1.00	1166246	1230
## 11982	5	5.50	144683	7640
## 11992	3	1.50	425145	1670
## 12212	2	1.00	167838	820
## 12255	2	1.75	227383	1650
## 12256	4	3.25	227383	3260
## 12392	3	2.25	188614	2140
## 12472	3	3.00	30894	3150
## 12487	4	2.50	9600	5000
## 12577	3	3.00	227818	5030
## 12582	5	4.75	29494	7070
## 12643	3	3.25	77418	4690
## 12686	3	3.50	19290	3720
## 12764	5	6.00	226512	6990
## 12816	4	3.75	1008414	5270
##	standardized.residuals			
## 6		-2.169095		
## 25		-2.464169		
## 115		2.791772		
## 178		-2.474876		
## 239		2.052786		
## 246		3.041611		
## 287		-2.574517		
## 295		-4.430179		
## 300		-3.615300		
## 341		-2.028765		
## 359		-2.904703		
## 385		3.131682		
## 396		2.776649		
## 475		2.031579		

## 482	-2.389959
## 508	-2.390140
## 528	-2.392852
## 661	2.341984
## 670	4.221666
## 679	-4.093951
## 784	-2.319091
## 811	5.014196
## 877	-2.567943
## 916	-3.350770
## 1009	-2.121478
## 1119	-3.193074
## 1142	-2.928763
## 1155	-3.099174
## 1207	-2.281290
## 1380	-2.180859
## 1442	2.176322
## 1492	-2.249218
## 1504	-3.385120
## 1550	-2.623192
## 1633	-2.027268
## 1650	-4.459701
## 1716	-2.953151
## 1745	2.832130
## 1870	2.204127
## 1962	3.608411
## 1963	2.254210
## 1964	3.560400
## 1976	4.821815
## 1977	5.103647
## 1978	4.985045
## 1979	5.083321
## 1980	5.011166
## 1981	4.792488
## 1982	4.914306
## 2022	2.721711
## 2099	2.277989
## 2137	2.973717
## 2157	2.507200
## 2257	-2.024753
## 2264	3.285043
## 2302	-2.510455
## 2360	2.350515
## 2361	3.801963
## 2469	2.657148
## 2604	-2.138436
## 2684	4.939193
## 2685	4.756145
## 2686	4.364336
## 2687	4.847880
## 2688	4.954693
## 2689	4.507731
## 2690	4.399695
## 2699	-2.524554

## 2708	3.403173
## 2709	3.185061
## 2710	3.277592
## 2717	3.851219
## 2742	-2.648810
## 2852	4.396888
## 2934	-2.951760
## 2937	-2.900290
## 3097	3.754328
## 3102	2.762757
## 3110	2.264796
## 3111	2.110718
## 3168	7.645600
## 3169	7.562029
## 3170	7.437659
## 3171	7.366998
## 3172	7.382018
## 3173	7.407897
## 3174	7.382054
## 3175	7.550566
## 3176	7.326036
## 3177	7.315829
## 3178	7.356953
## 3179	7.437913
## 3180	7.315571
## 3181	7.437967
## 3182	7.645552
## 3183	7.645639
## 3184	7.645639
## 3185	7.645547
## 3186	7.645600
## 3187	7.645600
## 3188	7.561683
## 3189	7.550634
## 3190	7.382230
## 3191	7.550661
## 3192	7.366971
## 3193	7.382271
## 3194	7.382298
## 3195	7.315773
## 3196	7.315660
## 3197	7.325423
## 3198	7.645455
## 3199	7.645463
## 3200	7.645639
## 3201	7.645639
## 3202	7.645639
## 3424	-2.714626
## 3464	7.481137
## 3465	7.368747
## 3466	7.246342
## 3467	7.246088
## 3468	7.481064
## 3469	7.368710

## 3470	7.246410
## 3471	7.368683
## 3472	7.297433
## 3473	7.312585
## 3474	7.480898
## 3475	7.297162
## 3476	7.334545
## 3477	7.297251
## 3478	7.576084
## 3479	7.576084
## 3480	7.576084
## 3481	7.576084
## 3482	7.576084
## 3483	7.576084
## 3484	7.492520
## 3485	7.246363
## 3486	7.368742
## 3487	7.246358
## 3488	7.312752
## 3489	7.297549
## 3490	7.353731
## 3491	7.343404
## 3492	7.409813
## 3493	7.246286
## 3494	7.481046
## 3495	7.576084
## 3496	7.576084
## 3497	7.576084
## 3523	2.886146
## 3810	-2.114331
## 3837	-2.447865
## 3918	-2.578212
## 3919	-2.383669
## 4055	2.801724
## 4056	4.496306
## 4285	-2.165407
## 4435	-2.307624
## 4571	-2.266753
## 4648	8.523399
## 4649	10.377699
## 4671	-2.625073
## 4695	5.073738
## 4696	3.098521
## 4821	2.917134
## 4834	2.485338
## 4840	-3.171840
## 4934	2.283064
## 5083	-4.009708
## 5491	2.301231
## 5494	2.276028
## 5495	2.280842
## 5496	2.301271
## 5497	2.341703
## 5498	2.037649

## 5549	-2.057601
## 5935	-3.064996
## 6055	-2.001481
## 6230	5.770388
## 6231	5.663625
## 6232	5.546179
## 6233	5.898064
## 6234	5.816190
## 6235	5.888332
## 6236	5.816167
## 6237	5.888344
## 6238	5.943818
## 6239	5.663645
## 6429	10.007608
## 6430	10.434894
## 6431	10.282021
## 6432	10.128252
## 6433	10.131055
## 6434	10.131028
## 6435	9.812642
## 6436	10.249534
## 6437	10.384111
## 6438	10.364011
## 6439	10.073809
## 6440	9.848995
## 6441	10.073883
## 6442	10.078406
## 6443	9.532674
## 6444	9.369598
## 6445	9.654582
## 6446	9.420092
## 6447	9.241900
## 6448	9.699904
## 6449	9.201090
## 6450	9.048833
## 6451	9.649630
## 6452	9.409939
## 6453	9.206615
## 6454	9.328668
## 6455	9.608477
## 6456	9.648686
## 6457	9.333821
## 6512	3.739823
## 6527	-2.338321
## 6634	2.094367
## 6739	-2.581460
## 6821	-2.512063
## 6931	-2.123275
## 6938	3.854013
## 6939	3.445945
## 6940	4.184658
## 6941	4.179072
## 6942	3.633586
## 6943	3.618164

## 6944	3.451195
## 6945	3.791733
## 6946	4.179521
## 6947	4.235346
## 6948	4.260474
## 7039	2.010750
## 7147	-2.538052
## 7167	6.105715
## 7210	4.090300
## 7211	4.595294
## 7389	-2.005088
## 7446	7.778132
## 7447	7.844188
## 7448	7.775474
## 7449	7.768260
## 7450	7.770804
## 7451	7.773343
## 7452	7.844120
## 7453	7.788202
## 7454	7.783096
## 7455	7.839037
## 7456	7.844197
## 7457	7.783256
## 7458	7.773123
## 7459	7.882476
## 7460	7.770804
## 7461	7.867211
## 7462	7.874843
## 7463	7.867207
## 7507	-2.584398
## 7649	-2.104840
## 7650	-2.104777
## 7791	-2.159119
## 7871	-3.153005
## 8119	-2.289467
## 8154	-2.079921
## 8232	-2.594661
## 8262	-4.102206
## 8320	-2.880790
## 8377	-5.674254
## 8457	2.215265
## 8458	3.521445
## 8535	-2.058259
## 8541	-2.034223
## 8698	2.440589
## 8710	4.557777
## 8763	2.400755
## 8887	5.304683
## 8911	2.160610
## 8946	2.160610
## 9215	2.600894
## 9293	-2.420684
## 9420	-2.017462
## 9453	-2.122675

```
## 9528      -4.280905
## 9546      -2.392214
## 9722       2.650307
## 10125     2.095178
## 10318     2.501370
## 10371     -2.188329
## 10418     -2.212801
## 10623     -2.046851
## 10707     2.169470
## 10723     2.123158
## 10787     3.183820
## 10958     -2.509562
## 10995     -2.385652
## 11165     2.480460
## 11289     4.146703
## 11413     3.697991
## 11558     5.075753
## 11586     2.813639
## 11728     3.257622
## 11758     -2.271830
## 11772     6.506286
## 11822     5.127549
## 11898     3.268641
## 11899     4.562752
## 11982     4.238087
## 11992    10.557864
## 12212     2.347121
## 12255     3.322548
## 12256     2.530938
## 12392     2.338030
## 12472     4.306578
## 12487     3.073246
## 12577     4.275691
## 12582     6.777127
## 12643     7.665506
## 12686     -2.146865
## 12764     3.986220
## 12816     2.177011
```

Investigate further by calculating the leverage, cooks distance, and covariance ratios. Comment on all cases that are problematic.

K. Based on the data, line 295 is the only problematic case since the leverage exceeds the average leverage amount. The covariance ratio also falls outside of the average boundry.

```
housing_df[housing_df$large.residual , c("cooks.distance", "leverage", "covariance.ratios")]
```

```
##      cooks.distance      leverage covariance.ratios
## 6      0.0003219879 0.0003420612      0.9989019
## 25     0.0011898691 0.0009788202      0.9990072
## 115    0.0037880501 0.0024242182      0.9997848
## 178    0.0011844718 0.0009659797      0.9989738
## 239    0.0006233065 0.0007390322      0.9994896
## 246    0.0008614624 0.0004653684      0.9972598
```


## 287	0.0008033656	0.0006056595	0.9984182
## 295	0.6830692450	0.1482237082	1.1655389
## 300	0.0070533441	0.0026909534	0.9980010
## 341	0.0003517824	0.0004271644	0.9992159
## 359	0.0044454247	0.0026274674	0.9997383
## 385	0.0033264346	0.0016930038	0.9982701
## 396	0.0016896828	0.0010946056	0.9984867
## 475	0.0001178536	0.0001427523	0.9989272
## 482	0.0031317813	0.0027339556	1.0009056
## 508	0.0026772302	0.0023377214	1.0005078
## 528	0.0014287234	0.0012460774	0.9994092
## 661	0.0040402732	0.0036695793	1.0019340
## 670	0.0021012086	0.0005891360	0.9940616
## 679	0.0156268687	0.0046402008	0.9985202
## 784	0.0016050049	0.0014899211	0.9997884
## 811	0.0045431943	0.0009026856	0.9915430
## 877	0.0015613197	0.0011824346	0.9990079
## 916	0.0055593886	0.0024696423	0.9984954
## 1009	0.0002607830	0.0002896318	0.9989289
## 1119	0.0038310064	0.0018752094	0.9983015
## 1142	0.0021696267	0.0012630999	0.9983180
## 1155	0.0185235733	0.0095506991	1.0062692
## 1207	0.0045867821	0.0043874045	1.0027658
## 1380	0.0006927134	0.0007276999	0.9992675
## 1442	0.0012764748	0.0013457086	0.9998936
## 1492	0.0009571485	0.0009450952	0.9993672
## 1504	0.0072558839	0.0031560179	0.9990930
## 1550	0.0012358643	0.0008972029	0.9986113
## 1633	0.0002862613	0.0003481441	0.9991392
## 1650	0.0249069302	0.0062225412	0.9988926
## 1716	0.0008819944	0.0005054114	0.9975055
## 1745	0.0014477690	0.0009016774	0.9981730
## 1870	0.0066909754	0.0068392101	1.0053767
## 1962	0.0029744138	0.0011408895	0.9964716
## 1963	0.0002357643	0.0002319309	0.9986456
## 1964	0.0018930113	0.0007461076	0.9962113
## 1976	0.0019736051	0.0004242529	0.9917992
## 1977	0.0011081664	0.0002126774	0.9905093
## 1978	0.0008115818	0.0001632650	0.9909222
## 1979	0.0011593927	0.0002242890	0.9906008
## 1980	0.0013513758	0.0002689997	0.9909262
## 1981	0.0033687873	0.0007328297	0.9922144
## 1982	0.0017513891	0.0003624690	0.9913901
## 2022	0.0010934821	0.0007375261	0.9982472
## 2099	0.0007955582	0.0007659595	0.9991374
## 2137	0.0003268155	0.0001847538	0.9971383
## 2157	0.0002351864	0.0001870351	0.9981330
## 2257	0.0001385184	0.0001689118	0.9989641
## 2264	0.0003892849	0.0001803338	0.9963782
## 2302	0.0107245886	0.0084365774	1.0064308
## 2360	0.0002708650	0.0002450702	0.9984865
## 2361	0.0049731284	0.0017172659	0.9964905
## 2469	0.0002562921	0.0001814659	0.9978268
## 2604	0.0005762901	0.0006297165	0.9992407

## 2684	0.0024344026	0.0004986936	0.9914305
## 2685	0.0012136644	0.0002681896	0.9918873
## 2686	0.0020010364	0.0005250016	0.9935238
## 2687	0.0012436596	0.0002645166	0.9915434
## 2688	0.0009143186	0.0001861886	0.9910614
## 2689	0.0018327631	0.0004507799	0.9929580
## 2690	0.0024165596	0.0006238085	0.9935022
## 2699	0.0311338112	0.0238425591	1.0222863
## 2708	0.0016972244	0.0007321887	0.9966220
## 2709	0.0016634813	0.0008192124	0.9972663
## 2710	0.0016976100	0.0007895052	0.9970046
## 2717	0.0037993554	0.0012791681	0.9959072
## 2742	0.0010161724	0.0007236379	0.9983854
## 2852	0.0278656415	0.0071553285	1.0000476
## 2934	0.0008811637	0.0005054114	0.9975087
## 2937	0.0008507018	0.0005054114	0.9976256
## 3097	0.0032374787	0.0011471339	0.9960612
## 3102	0.0009313523	0.0006097246	0.9980321
## 3110	0.0004050927	0.0003947253	0.9987897
## 3111	0.0005985973	0.0006713558	0.9993282
## 3168	0.0060525798	0.0005174431	0.9783645
## 3169	0.0038543534	0.0003368981	0.9786735
## 3170	0.0027294014	0.0002466367	0.9792983
## 3171	0.0022190483	0.0002043934	0.9796570
## 3172	0.0023177687	0.0002126168	0.9795803
## 3173	0.0014464610	0.0001317740	0.9793548
## 3174	0.0023178191	0.0002126194	0.9795801
## 3175	0.0019779541	0.0001734410	0.9785797
## 3176	0.0019764616	0.0001840943	0.9798673
## 3177	0.0019212439	0.0001794516	0.9799200
## 3178	0.0013524890	0.0001249262	0.9796357
## 3179	0.0027276367	0.0002464605	0.9792967
## 3180	0.0019194083	0.0001792929	0.9799213
## 3181	0.0027273345	0.0002464296	0.9792963
## 3182	0.0060503319	0.0005172575	0.9783646
## 3183	0.0060544277	0.0005175957	0.9783645
## 3184	0.0060544277	0.0005175957	0.9783645
## 3185	0.0060500694	0.0005172358	0.9783646
## 3186	0.0060525798	0.0005174431	0.9783645
## 3187	0.0060525798	0.0005174431	0.9783645
## 3188	0.0038496484	0.0003365178	0.9786751
## 3189	0.0019782572	0.0001734645	0.9785793
## 3190	0.0023182201	0.0002126460	0.9795792
## 3191	0.0019783910	0.0001734750	0.9785792
## 3192	0.0022189152	0.0002043826	0.9796571
## 3193	0.0023183501	0.0002126556	0.9795789
## 3194	0.0023184445	0.0002126626	0.9795788
## 3195	0.0019208004	0.0001794130	0.9799202
## 3196	0.0019199822	0.0001793421	0.9799208
## 3197	0.0019735017	0.0001838494	0.9798705
## 3198	0.0060458335	0.0005168862	0.9783648
## 3199	0.0060462022	0.0005169166	0.9783648
## 3200	0.0060544277	0.0005175957	0.9783645
## 3201	0.0060544277	0.0005175957	0.9783645

## 3202	0.0060544277	0.0005175957	0.9783645
## 3424	0.0005655748	0.0003835955	0.9979087
## 3464	0.0019422097	0.0001734825	0.9789787
## 3465	0.0026752771	0.0002462887	0.9796882
## 3466	0.0018853452	0.0001794921	0.9803072
## 3467	0.0018834587	0.0001793251	0.9803085
## 3468	0.0019418620	0.0001734548	0.9789790
## 3469	0.0026753819	0.0002463009	0.9796884
## 3470	0.0018859354	0.0001795449	0.9803069
## 3471	0.0026754620	0.0002463100	0.9796886
## 3472	0.0021772473	0.0002043849	0.9800473
## 3473	0.0022744921	0.0002126279	0.9799706
## 3474	0.0019412363	0.0001734066	0.9789799
## 3475	0.0021762626	0.0002043076	0.9800487
## 3476	0.0027702504	0.0002574132	0.9798915
## 3477	0.0021765222	0.0002043270	0.9800482
## 3478	0.0059447711	0.0005175957	0.9787691
## 3479	0.0059447711	0.0005175957	0.9787691
## 3480	0.0059447711	0.0005175957	0.9787691
## 3481	0.0059447711	0.0005175957	0.9787691
## 3482	0.0059447711	0.0005175957	0.9787691
## 3483	0.0059447587	0.0005175947	0.9787691
## 3484	0.0037843708	0.0003369469	0.9790736
## 3485	0.0018855220	0.0001795079	0.9803071
## 3486	0.0026752910	0.0002462903	0.9796882
## 3487	0.0018854859	0.0001795047	0.9803071
## 3488	0.0022749608	0.0002126620	0.9799697
## 3489	0.0021778451	0.0002044345	0.9800467
## 3490	0.0014579459	0.0001347837	0.9796635
## 3491	0.0025229094	0.0002338704	0.9798187
## 3492	0.0016318200	0.0001485810	0.9793604
## 3493	0.0018848895	0.0001794515	0.9803075
## 3494	0.0019417835	0.0001734486	0.9789791
## 3495	0.0059447587	0.0005175947	0.9787691
## 3496	0.0059447587	0.0005175947	0.9787691
## 3497	0.0059447587	0.0005175947	0.9787691
## 3523	0.0010919963	0.0006550435	0.9978068
## 3810	0.0003154425	0.0003526884	0.9990037
## 3837	0.0009313964	0.0007765908	0.9988361
## 3918	0.0024976527	0.0018752094	0.9996807
## 3919	0.0014371717	0.0012630999	0.9994433
## 4055	0.0013437944	0.0008552251	0.9981931
## 4056	0.0031260006	0.0007725234	0.9933176
## 4285	0.0004287551	0.0004569849	0.9990230
## 4435	0.0005603310	0.0005258430	0.9988446
## 4571	0.0010183176	0.0009899537	0.9993813
## 4648	0.1477412726	0.0100658982	0.9823376
## 4649	0.9882328762	0.0438677257	1.0032039
## 4671	0.0067388895	0.0048658344	1.0025899
## 4695	0.0015582268	0.0003025612	0.9907159
## 4696	0.0018093677	0.0009414104	0.9975993
## 4821	0.0013186234	0.0007741797	0.9978559
## 4834	0.0009117984	0.0007375261	0.9987253
## 4840	0.0194023950	0.0095506991	1.0060908

## 4934	0.0004947700	0.0004743852	0.9988370
## 5083	0.1013904460	0.0305674475	1.0254979
## 5491	0.0001447041	0.0001366065	0.9984672
## 5494	0.0001495460	0.0001443202	0.9985197
## 5495	0.0001198809	0.0001152072	0.9984821
## 5496	0.0001448868	0.0001367740	0.9984673
## 5497	0.0003993209	0.0003639738	0.9986213
## 5498	0.0002195736	0.0002643484	0.9990391
## 5549	0.0005463996	0.0006448787	0.9993877
## 5935	0.0029661752	0.0015762385	0.9983139
## 6055	0.0008511138	0.0010611908	0.9998928
## 6230	0.0012454654	0.0001869865	0.9876893
## 6231	0.0012737730	0.0001985120	0.9881708
## 6232	0.0016752806	0.0002722397	0.9887509
## 6233	0.0009131483	0.0001312307	0.9870608
## 6234	0.0027550030	0.0004070410	0.9877024
## 6235	0.0024535747	0.0003536967	0.9873247
## 6236	0.0027546193	0.0004069875	0.9877025
## 6237	0.0024539196	0.0003537451	0.9873246
## 6238	0.0008646876	0.0001223616	0.9868436
## 6239	0.0012744471	0.0001986157	0.9881708
## 6429	0.0040611931	0.0002027099	0.9622314
## 6430	0.0028399546	0.0001303913	0.9588731
## 6431	0.0025703674	0.0001215501	0.9600560
## 6432	0.0023615627	0.0001150934	0.9612316
## 6433	0.0084892631	0.0004133816	0.9614970
## 6434	0.0084868729	0.0004132674	0.9614971
## 6435	0.0086901806	0.0004510565	0.9639288
## 6436	0.0024923415	0.0001186090	0.9603042
## 6437	0.0026693437	0.0001237605	0.9592643
## 6438	0.0025965552	0.0001208535	0.9594184
## 6439	0.0035088048	0.0001728489	0.9617015
## 6440	0.0075722501	0.0003901590	0.9636002
## 6441	0.0100280980	0.0004938331	0.9620098
## 6442	0.0025731926	0.0001266495	0.9616222
## 6443	0.0097385388	0.0005355520	0.9660573
## 6444	0.0076285830	0.0004342935	0.9671259
## 6445	0.0089226026	0.0004783952	0.9651176
## 6446	0.0058042720	0.0003269380	0.9666630
## 6447	0.0065750256	0.0003847493	0.9679781
## 6448	0.0022848771	0.0001214074	0.9644416
## 6449	0.0055596713	0.0003282449	0.9682086
## 6450	0.0057091361	0.0003485009	0.9692816
## 6451	0.0088334952	0.0004741058	0.9651496
## 6452	0.0057131500	0.0003225016	0.9667310
## 6453	0.0057063012	0.0003364950	0.9681780
## 6454	0.0072343482	0.0004154795	0.9673975
## 6455	0.0071843237	0.0003889357	0.9653670
## 6456	0.0022135914	0.0001188720	0.9648136
## 6457	0.0060000655	0.0003442370	0.9672922
## 6512	0.0022613219	0.0008077545	0.9957650
## 6527	0.0006873714	0.0006281744	0.9988915
## 6634	0.0003656347	0.0004166112	0.9991002
## 6739	0.0016507933	0.0012370698	0.9990355

## 6821	0.0008969290	0.0007101639	0.9986460
## 6931	0.0004497079	0.0004985074	0.9991347
## 6938	0.0011007655	0.0003704056	0.9949935
## 6939	0.0012753414	0.0005367186	0.9963135
## 6940	0.0007823969	0.0002233470	0.9938183
## 6941	0.0007799793	0.0002232525	0.9938363
## 6942	0.0011860062	0.0004489430	0.9957111
## 6943	0.0013408590	0.0005118635	0.9958171
## 6944	0.0013554086	0.0005686619	0.9963313
## 6945	0.0008857915	0.0003079585	0.9951158
## 6946	0.0007897116	0.0002259890	0.9938375
## 6947	0.0006999035	0.0001950498	0.9936250
## 6948	0.0006585838	0.0001813788	0.9935289
## 7039	0.0008229204	0.0010166469	0.9998338
## 7147	0.0011228731	0.0008708060	0.9987556
## 7167	0.0046049858	0.0006172443	0.9865816
## 7210	0.0033953923	0.0010136992	0.9949069
## 7211	0.0030560496	0.0007230853	0.9929204
## 7389	0.0023376148	0.0028987810	1.0017299
## 7446	0.0014637780	0.0001209601	0.9771960
## 7447	0.0015386249	0.0001250122	0.9768061
## 7448	0.0014581478	0.0001205773	0.9772114
## 7449	0.0015032122	0.0001245342	0.9772581
## 7450	0.0015004397	0.0001242232	0.9772427
## 7451	0.0014975740	0.0001239050	0.9772273
## 7452	0.0015365304	0.0001248442	0.9768064
## 7453	0.0014659942	0.0001208302	0.9771361
## 7454	0.0014619179	0.0001206524	0.9771662
## 7455	0.0015274146	0.0001242646	0.9768362
## 7456	0.0015389347	0.0001250371	0.9768061
## 7457	0.0014679886	0.0001211483	0.9771658
## 7458	0.0014872726	0.0001230597	0.9772278
## 7459	0.0016288070	0.0001310561	0.9765823
## 7460	0.0015004397	0.0001242232	0.9772427
## 7461	0.0015906297	0.0001284818	0.9766715
## 7462	0.0016090630	0.0001297188	0.9766268
## 7463	0.0015905014	0.0001284716	0.9766715
## 7507	0.0019932400	0.0014899211	0.9992826
## 7649	0.0007513199	0.0008472052	0.9995137
## 7650	0.0007515725	0.0008475406	0.9995141
## 7791	0.0018467971	0.0019768625	1.0005549
## 7871	0.0062464964	0.0031318071	0.9996588
## 8119	0.0007450164	0.0007101639	0.9990613
## 8154	0.0002460322	0.0002842790	0.9989914
## 8232	0.0020091021	0.0014899211	0.9992619
## 8262	0.0364288334	0.0107079122	1.0046180
## 8320	0.0081084346	0.0048614669	1.0020365
## 8377	0.4309358879	0.0627238474	1.0540418
## 8457	0.0002276883	0.0002319309	0.9987132
## 8458	0.0018518144	0.0007461076	0.9963183
## 8535	0.0003967284	0.0004680153	0.9992098
## 8541	0.0002882288	0.0003481441	0.9991283
## 8698	0.0002121660	0.0001780652	0.9982520
## 8710	0.0027579977	0.0006633908	0.9929938

## 8763	0.0005156776	0.0004471553	0.9985956
## 8887	0.0119755803	0.0021233645	0.9915972
## 8911	0.0012676105	0.0013558568	0.9999302
## 8946	0.0012676105	0.0013558568	0.9999302
## 9215	0.0041687533	0.0030718123	1.0008349
## 9293	0.0014621525	0.0012460774	0.9993571
## 9420	0.0061700417	0.0075226038	1.0063774
## 9453	0.0004556817	0.0005054114	0.9991426
## 9528	0.0097687444	0.0026581628	0.9959285
## 9546	0.0008895283	0.0007765908	0.9989408
## 9722	0.0002386507	0.0001698504	0.9978293
## 10125	0.0019947489	0.0022668926	1.0009517
## 10318	0.0278515700	0.0217722761	1.0201690
## 10371	0.0013015228	0.0013570841	0.9998846
## 10418	0.0018453822	0.0018808495	1.0003674
## 10623	0.0004481403	0.0005345388	0.9992946
## 10707	0.0001107521	0.0001176424	0.9986771
## 10723	0.0005530879	0.0006131027	0.9992494
## 10787	0.0012454092	0.0006139295	0.9970645
## 10958	0.0006456080	0.0005122955	0.9984532
## 10995	0.0042389855	0.0037102518	1.0018945
## 11165	0.0010495342	0.0008521811	0.9988493
## 11289	0.0006242666	0.0001814912	0.9938990
## 11413	0.0022295657	0.0008145254	0.9958924
## 11558	0.0024055027	0.0004666295	0.9908706
## 11586	0.0001850899	0.0001168867	0.9974301
## 11728	0.0099609648	0.0046712882	1.0009438
## 11758	0.0014027427	0.0013570841	0.9997398
## 11772	0.0201404015	0.0023732306	0.9863727
## 11822	0.0032857839	0.0006244793	0.9908231
## 11898	0.0760698716	0.0343760691	1.0317062
## 11899	0.1490091795	0.0345508547	1.0278300
## 11982	0.0088936020	0.0024696423	0.9958817
## 11992	0.1019543278	0.0045524161	0.9621581
## 12212	0.0012160382	0.0011024709	0.9993497
## 12255	0.0031314656	0.0014163140	0.9975153
## 12256	0.0015887144	0.0012385528	0.9991374
## 12392	0.0009307557	0.0008506209	0.9991144
## 12472	0.0008656718	0.0002333228	0.9934277
## 12487	0.0024170684	0.0012779354	0.9979961
## 12577	0.0063660162	0.0017380796	0.9950278
## 12582	0.0174415408	0.0018951331	0.9845174
## 12643	0.0098110068	0.0008341412	0.9785581
## 12686	0.0004411970	0.0004783937	0.9990754
## 12764	0.0094720864	0.0029716702	0.9971870
## 12816	0.0231517743	0.0238425591	1.0229363

Perform the necessary calculations to assess the assumption of independence and state if the condition is met or not.

L. The condition is not met since the values are below the statistic value is below 1 and the p-value isn't above 0.05.

```
##{r} ##install.packages("car") ##library(car) ##
```

```
library(car)
```

```
## Warning: package 'car' was built under R version 4.0.2
```

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

```
##
```

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

```
## The following object is masked from 'package:boot':
```

```
##
```

```
##      logit
```

```
durbinWatsonTest(saleprice_lm)
```

```
## lag Autocorrelation D-W Statistic p-value
```

```
## 1 0.7309992 0.5379977 0
```

```
## Alternative hypothesis: rho != 0
```

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

M. Based on our data, there is no multicollinearity.

```
vif(saleprice_lm)
```

```
##          sq_ft_lot square_feet_total_living      bedrooms
##          1.085244          2.706329          1.551002
##          total_bath
##          2.335764
```

```
1/vif(saleprice_lm)
```

```
##          sq_ft_lot square_feet_total_living      bedrooms
##          0.9214520          0.3695042          0.6447445
##          total_bath
##          0.4281254
```

```
mean(vif(saleprice_lm))
```

```
## [1] 1.919585
```

N. 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.

The first plot is the residuals vs. the fitted values. This plot is used to show how far away from the average the residuals are.

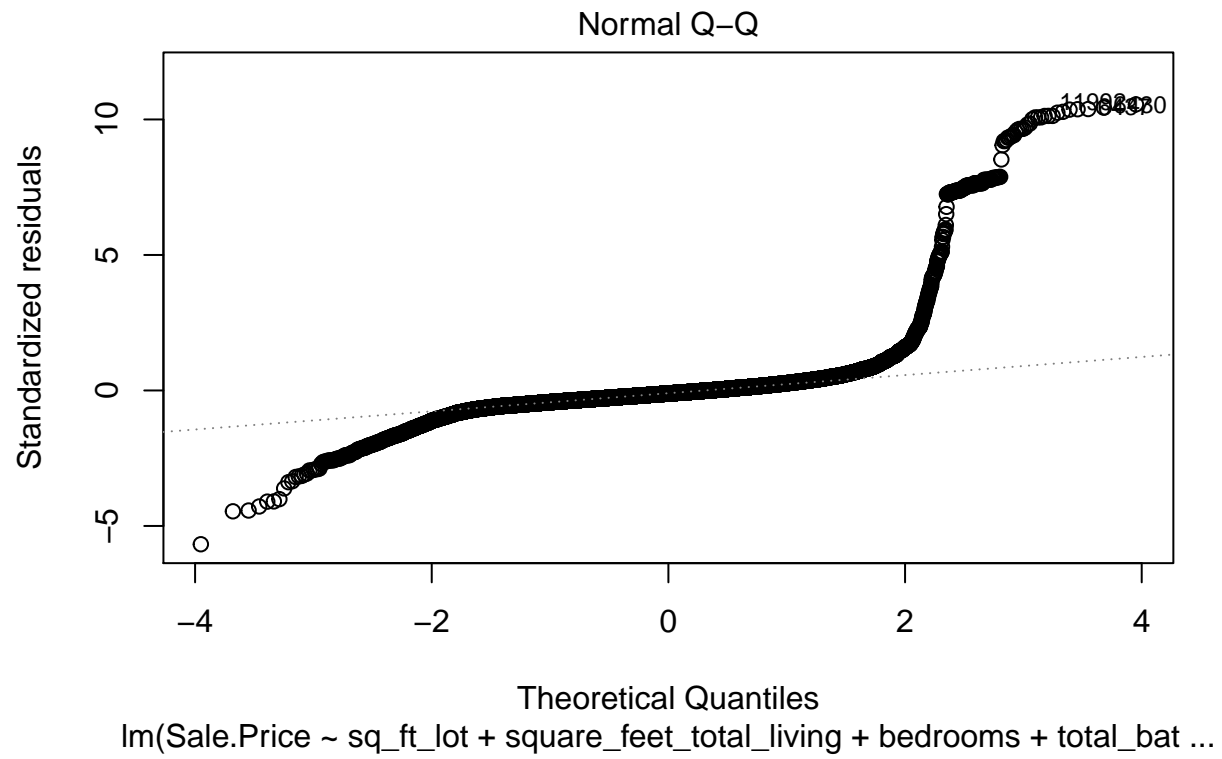
The second plot is the Q-Q plot. This plot illustrates probability distributions. My regression model doesn't appear to provide a normal Q-Q plot since the line tails off at the end.

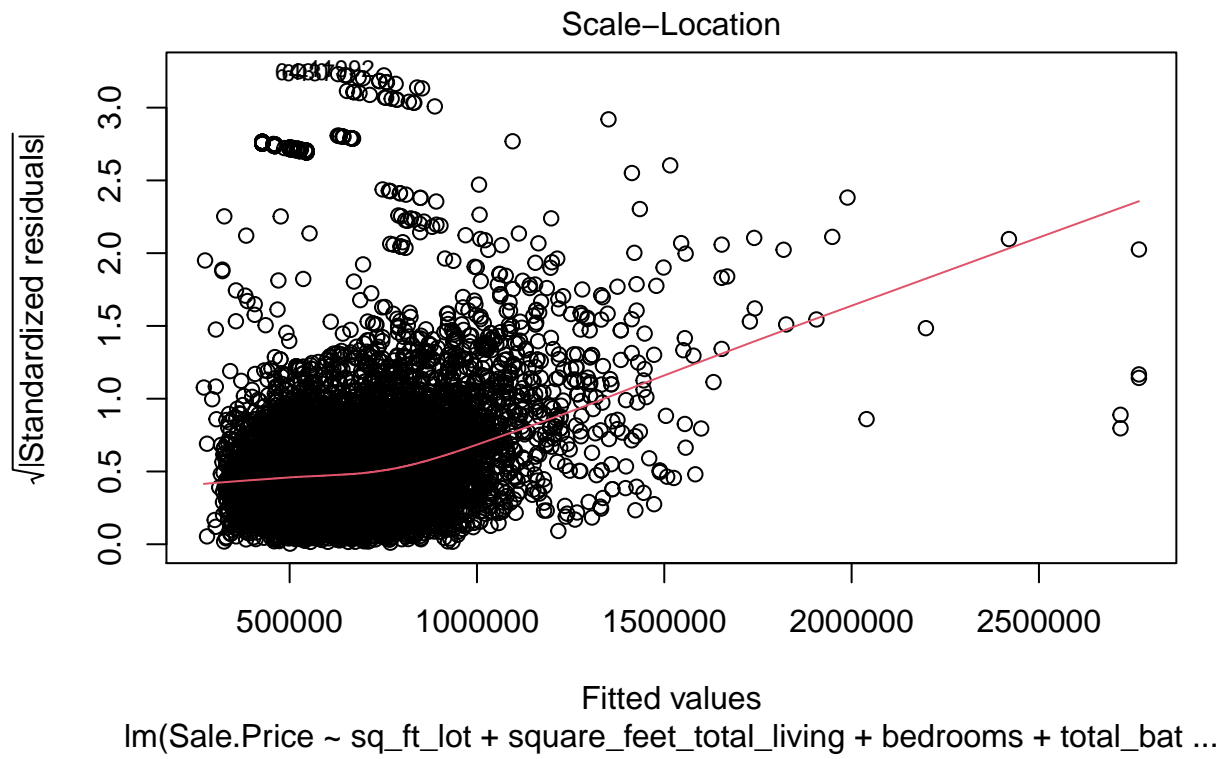
The third plot is a Scale-Location plot. This plot shows if residuals are spread equally. Mine appear to be fairly concentrated to the left.

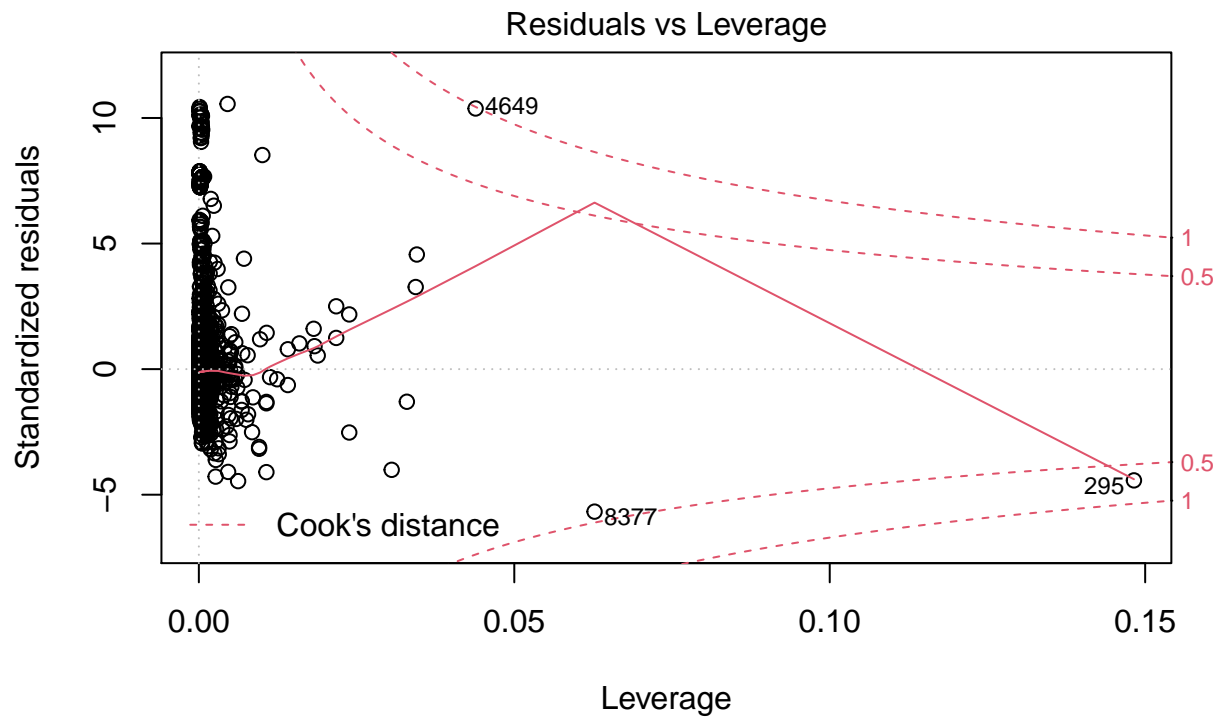
The fourth plot is the Residuals vs. Leverage plot. This plot is used to identify outliers using the Cook's distance lines. There are 3 cases identified to be anomalies that are impacting the data (8377, 295, and 4649).

```
plot(saleprice_lm)
```



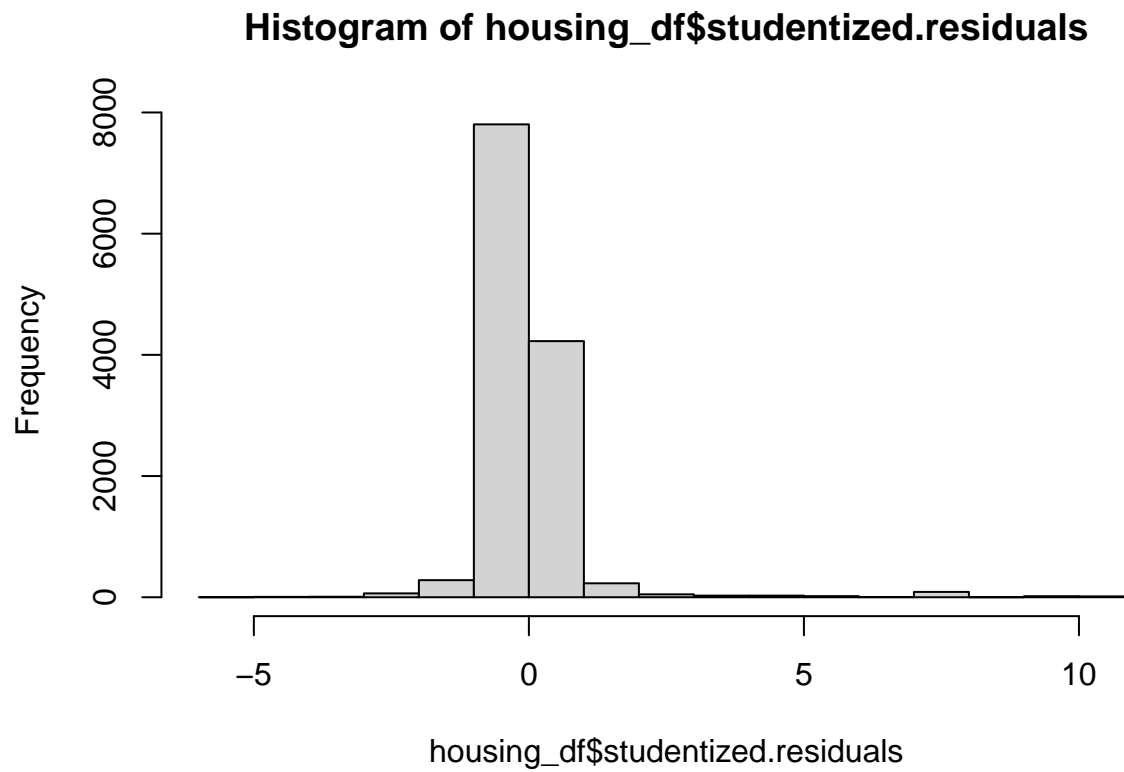






lm(Sale.Price ~ sq_ft_lot + square_feet_total_living + bedrooms + total_bat ...

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



Is the regression model unbiased? If an unbiased regression model, what does this tell us about the sample vs. the entire population?

O. Yes, the regression model is unbiased. This tells us that the sample is an accurate representation of the true population.