

## 6.1 Housing Data

Anna Harvey

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

n/a

b. 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.

```
sale_sqft <- lm(survey$SalePrice~survey$sq_ft_lot)
sale_vars <- lm(survey$SalePrice~survey$square_feet_total_living + survey$year_built)
```

I chose the square feet of the living space and the year the house was built as additional predictors. I chose these because the amount of living space and how old the house is are likely to determine the sale price of the house.

c. 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?

```
summary(sale_sqft)
```

```
##
## Call:
## lm(formula = survey$SalePrice ~ survey$sq_ft_lot)
##
## 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 ***
## survey$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(sale_vars)
```

```
##
## Call:
## lm(formula = survey$SalePrice ~ survey$square_feet_total_living +
##     survey$year_built)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1719467  -121308   -42621    44230   3916857
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -5.114e+06  3.808e+05  -13.43  <2e-16 ***
## survey$square_feet_total_living  1.714e+02  3.346e+00   51.24  <2e-16 ***
## survey$year_built      2.679e+03  1.923e+02   13.93  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 357500 on 12862 degrees of freedom
## Multiple R-squared:  0.2184, Adjusted R-squared:  0.2183
## F-statistic: 1797 on 2 and 12862 DF,  p-value: < 2.2e-16
```

Sale Price vs Sq\_ft\_lot  $R^2 = 0.01435$  Adjusted  $R^2 = 0.01428$

Sale Price vs Sq Ft Living vs Year Built  $R^2 = 0.2184$  Adjusted  $R^2 = 0.2183$

$R^2$  and adjusted  $R^2$  increases with the addition of more variables. This tells me that the addition of my choice of variables increased the variation of the output value. The  $R^2$  and adjusted  $R^2$  for each variable are very close to each other. This tells me that the variables are significant and do affect the output variable.

**d. 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?**

```
lm.beta(sale_vars)
```

```
## survey$square_feet_total_living    survey$year_built
##              0.4196286              0.1140856
```

Standardized beta for Sq Ft Living Space = 0.4196286 Standardized beta for Year Built = 0.1140856

This tells me that the square footage of the living space plays a more important role in determining the sale price than the year the house was built.

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

```
confint(sale_vars)
```

```
##                2.5 %      97.5 %
## (Intercept)    -5860366.2970 -4367673.7301
## survey$square_feet_total_living    164.8777    177.9934
## survey$year_built    2302.1248    3056.0179
```

The variation in the confidence intervals for both the Living Space and Year Built are fairly close to each other, indicating they are likely good representations of the general trend. Living Space is very close together which follows the information from the beta values indicating that it is a more important variable than Year Built for determining the sale price.

f. 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.

```
anova(sale_sqft, sale_vars)
```

```
## Analysis of Variance Table
##
## Model 1: survey$SalePrice ~ survey$sq_ft_lot
## Model 2: survey$SalePrice ~ survey$square_feet_total_living + survey$year_built
##   Res.Df      RSS Df Sum of Sq    F    Pr(>F)
## 1  12863 2.0734e+15
## 2  12862 1.6441e+15  1 4.2931e+14 3358.7 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

F = 3358.7 This does seem to indicate that the addition of the variables caused a significant improvement to the model.

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

```
sale_varsdf <- data.frame(survey$SalePrice, survey$square_feet_total_living, survey$year_built)
sale_varsdf$residuals <- resid(sale_vars)
sale_varsdf$rstandard <- rstandard(sale_vars)
sale_varsdf$cooks <- cooks.distance(sale_vars)
sale_varsdf$covratio <- covratio(sale_vars)
sale_varsdf$leverage <- hatvalues(sale_vars)
```

h. Calculate the standardized residuals using the appropriate command, specifying those that are  $\pm 2$ , storing the results of large residuals in a variable you create.

```
sale_varsdf$large_residuals <- sale_varsdf$rstandard > 2 | sale_varsdf$rstandard < -2  
large_residuals <- sale_varsdf$rstandard > 2 | sale_varsdf$rstandard < -2
```

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

```
sum(large_residuals)
```

```
## [1] 327
```

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

```
sale_varsdf[sale_varsdf$large_residuals == TRUE, c("survey.SalePrice", "survey.square_feet_total_living", "survey.year_built")]
```

##	survey.SalePrice	survey.square_feet_total_living	survey.year_built
## 6	184667	4160	2005
## 25	265000	4920	2007
## 115	1390000	660	1955
## 178	390000	5800	2008
## 239	1588359	3360	2005
## 246	1450000	900	1918
## 287	163000	4710	2014
## 295	270000	5060	2016
## 300	200000	6880	2008
## 341	300000	4490	2008
## 359	187000	5140	2008
## 385	2500000	6310	2005
## 396	2169000	5080	2005
## 475	1534000	3320	1963
## 482	555000	6380	2007
## 508	65000	3700	2010
## 528	435000	5830	2014
## 661	2569000	8090	2006
## 670	2583000	4710	1992
## 679	350000	8490	2008
## 784	325000	5150	2015
## 811	3000000	5270	2001
## 853	1595000	3470	1985
## 877	157000	4640	2008
## 916	450000	7640	2007
## 1009	175000	4040	2005
## 1119	279150	6340	2007
## 1142	279150	5980	2007

## 1155	32000	4740	2007
## 1305	1316000	2800	1929
## 1345	233333	3990	2008
## 1380	80000	3660	2008
## 1442	1085000	340	1954
## 1492	285000	4610	2007
## 1504	452800	7780	2008
## 1550	276450	5380	2007
## 1633	5000	3130	2014
## 1650	349999	9360	2009
## 1716	1000	4610	2015
## 1745	2235000	5360	2006
## 1870	2988000	10630	2003
## 1962	2493000	5300	2006
## 1963	1600000	3090	1978
## 1964	1600000	550	1979
## 1976	2625000	3830	2009
## 1977	2625000	3330	2009
## 1978	2625000	3480	2008
## 1979	2625000	3370	2008
## 1980	2625000	3350	2008
## 1981	2625000	4080	2008
## 1982	2625000	3690	2009
## 2020	1862000	5310	1985
## 2022	1384950	900	1940
## 2099	1975000	5020	2004
## 2137	1787000	2820	1972
## 2157	1678000	3130	1965
## 2257	20000	2940	2015
## 2264	1651000	1540	1938
## 2302	450000	6340	2009
## 2360	1640000	3200	1968
## 2361	1640000	310	1964
## 2469	1710000	2980	1980
## 2604	40191	3060	2008
## 2684	2590000	3410	2008
## 2685	2590000	3660	2010
## 2686	2590000	4500	2008
## 2687	2590000	3480	2010
## 2688	2590000	3310	2008
## 2689	2590000	4340	2008
## 2690	2590000	4390	2009
## 2699	379950	5270	2015
## 2708	2300000	4710	2015
## 2709	2300000	5100	2016
## 2710	2300000	4910	2015
## 2717	2598000	5390	2007
## 2742	10570	3950	2013
## 2852	3995000	11810	2000
## 2934	1500	4610	2015
## 2937	20000	4610	2015
## 3097	2549000	5320	2007
## 3102	2080000	4790	2008
## 3110	1250000	1290	1961

## 3111	1250000	1920	1961
## 3168	3175000	1290	2008
## 3169	3175000	1290	2008
## 3170	3175000	1600	2009
## 3171	3175000	1740	2008
## 3172	3175000	1710	2008
## 3173	3175000	1740	2008
## 3174	3175000	1710	2008
## 3175	3175000	1460	2008
## 3176	3175000	1820	2009
## 3177	3175000	1840	2008
## 3178	3175000	1840	2008
## 3179	3175000	1600	2008
## 3180	3175000	1840	2008
## 3181	3175000	1600	2008
## 3182	3175000	1290	2008
## 3183	3175000	1290	2008
## 3184	3175000	1290	2008
## 3185	3175000	1290	2008
## 3186	3175000	1290	2008
## 3187	3175000	1290	2008
## 3188	3175000	1290	2008
## 3189	3175000	1460	2008
## 3190	3175000	1710	2008
## 3191	3175000	1460	2008
## 3192	3175000	1740	2008
## 3193	3175000	1710	2008
## 3194	3175000	1710	2008
## 3195	3175000	1840	2008
## 3196	3175000	1840	2008
## 3197	3175000	1820	2009
## 3198	3175000	1290	2008
## 3199	3175000	1290	2008
## 3200	3175000	1290	2008
## 3201	3175000	1290	2008
## 3202	3175000	1290	2008
## 3260	1750000	4604	1999
## 3424	20146	4290	2008
## 3464	3150000	1460	2008
## 3465	3150000	1600	2009
## 3466	3150000	1840	2008
## 3467	3150000	1840	2008
## 3468	3150000	1460	2008
## 3469	3150000	1600	2008
## 3470	3150000	1840	2008
## 3471	3150000	1600	2009
## 3472	3150000	1740	2008
## 3473	3150000	1710	2008
## 3474	3150000	1460	2008
## 3475	3150000	1740	2008
## 3476	3150000	1830	2008
## 3477	3150000	1740	2008
## 3478	3150000	1290	2008
## 3479	3150000	1290	2008

## 3480	3150000	1290	2008
## 3481	3150000	1290	2008
## 3482	3150000	1290	2008
## 3483	3150000	1290	2008
## 3484	3150000	1290	2008
## 3485	3150000	1840	2010
## 3486	3150000	1600	2008
## 3487	3150000	1840	2009
## 3488	3150000	1710	2008
## 3489	3150000	1740	2008
## 3490	3150000	1710	2008
## 3491	3150000	1840	2008
## 3492	3150000	1600	2009
## 3493	3150000	1840	2008
## 3494	3150000	1460	2008
## 3495	3150000	1290	2008
## 3496	3150000	1290	2008
## 3497	3150000	1290	2008
## 3523	2000000	3980	1955
## 3837	130000	4480	2014
## 3918	500000	6340	2007
## 3919	475000	5980	2007
## 4055	2000000	4110	1991
## 4056	2000000	900	1937
## 4285	155026	4180	2003
## 4435	188750	4330	2011
## 4648	4400000	5790	1999
## 4649	4400000	2410	1935
## 4671	800000	7810	2011
## 4695	2300000	1430	2003
## 4696	2300000	5330	2005
## 4740	1448000	2900	1974
## 4750	2300000	7640	2007
## 4821	1430000	890	1999
## 4834	1300000	900	1940
## 4840	6000	4740	2007
## 4934	1780000	4210	1959
## 5083	698	5830	1969
## 5491	1550000	2960	2011
## 5494	1550000	3010	2011
## 5495	1550000	2960	2011
## 5496	1550000	2960	2011
## 5497	1550000	2950	2011
## 5549	241000	4130	2013
## 5935	180000	5640	2012
## 6055	130000	3500	2013
## 6230	2885000	3320	2012
## 6231	2885000	3530	2014
## 6232	2885000	3720	2013
## 6233	2885000	3150	2013
## 6234	2885000	3340	2012
## 6235	2885000	3320	2012
## 6236	2885000	3340	2012
## 6237	2885000	3320	2013

## 6238	2885000	3060	2013
## 6239	2885000	3530	2014
## 6429	4380542	3290	2012
## 6430	4380542	2450	2010
## 6431	4380542	2750	2012
## 6432	4380542	3010	2012
## 6433	4380542	3200	2010
## 6434	4380542	3200	2012
## 6435	4380542	3620	2012
## 6436	4380542	2810	2012
## 6437	4380542	2550	2010
## 6438	4380542	2440	2011
## 6439	4380542	3160	2011
## 6440	4380542	3400	2012
## 6441	4380542	2960	2012
## 6442	4380542	3110	2012
## 6443	4140203	2900	2012
## 6444	4140203	3220	2012
## 6445	4140203	2510	2013
## 6446	4140203	2970	2013
## 6447	4140203	3470	2012
## 6448	4140203	2580	2011
## 6449	4140203	3590	2012
## 6450	4140203	3890	2012
## 6451	4140203	2520	2012
## 6452	4140203	2990	2013
## 6453	4140203	3620	2013
## 6454	4140203	3300	2012
## 6455	4140203	2830	2012
## 6456	4140203	2680	2012
## 6457	4140203	3330	2012
## 6512	2500000	5040	1990
## 6527	275000	5050	2012
## 6739	50000	3850	2013
## 6796	226610	4000	2014
## 6821	20000	4200	1955
## 6931	149650	4000	2004
## 6938	2300000	3970	2013
## 6939	2300000	4610	2013
## 6940	2300000	3360	2013
## 6941	2300000	3370	2013
## 6942	2300000	4390	2013
## 6943	2300000	4380	2013
## 6944	2300000	4790	2013
## 6945	2300000	3970	2013
## 6946	2300000	3370	2013
## 6947	2300000	3260	2013
## 6948	2300000	3210	2013
## 7039	1920000	5320	1976
## 7147	254000	5370	2014
## 7167	3200000	4460	1975
## 7210	2500000	4460	1978
## 7211	2500000	3640	1982
## 7446	3462000	2650	2013



## 7447	3462000	2520	2013
## 7448	3462000	2655	2013
## 7449	3462000	2520	2013
## 7450	3462000	2515	2013
## 7451	3462000	2510	2013
## 7452	3462000	2520	2013
## 7453	3462000	2630	2012
## 7454	3462000	2640	2012
## 7455	3462000	2530	2012
## 7456	3462000	2520	2013
## 7457	3462000	2640	2013
## 7458	3462000	2510	2013
## 7459	3462000	2445	2013
## 7460	3462000	2515	2013
## 7461	3462000	2475	2013
## 7462	3462000	2460	2013
## 7463	3462000	2475	2013
## 7507	229687	5150	2015
## 7649	280000	4350	2014
## 7650	280000	4350	2014
## 7791	525400	5620	2015
## 7871	347000	6650	2015
## 8119	100000	4200	1955
## 8154	192000	3960	2014
## 8232	226000	5150	2015
## 8262	1300000	13540	1999
## 8320	302500	5820	2016
## 8377	14000	8750	1996
## 8457	1586000	3090	1978
## 8458	1586000	550	1979
## 8535	245000	4420	2014
## 8541	2500	3130	2014
## 8698	1700000	3380	2014
## 8710	2750000	4950	1996
## 8763	1650000	3460	1972
## 8887	3340000	6360	2008
## 8911	2160200	6280	2003
## 8946	2160200	6280	2003
## 9215	1401322	1650	1954
## 9293	425000	5830	2014
## 9420	320000	4800	2015
## 9453	299500	4610	2015
## 9528	8000	7110	2006
## 9546	150000	4480	2014
## 9722	1700000	2900	1974
## 10125	2200000	6680	1998
## 10318	2140000	5150	1997
## 10371	299000	4930	2016
## 10418	325000	4770	2003
## 10623	321000	4570	2016
## 10707	1442500	2630	1987
## 10723	1850000	4790	2008
## 10741	285000	4200	2015
## 10787	2200000	4640	1952

## 10844	1200000	1800	1918
## 10958	18000	4010	1997
## 10995	1050000	9070	2005
## 11165	1890000	4140	1995
## 11289	2300000	3310	1982
## 11413	2025000	2610	1948
## 11558	2150000	730	1952
## 11586	1700000	2770	1938
## 11728	1840000	2280	1981
## 11758	269000	4930	2016
## 11772	3750000	6600	2001
## 11822	2850000	4240	2005
## 11898	2165000	3690	1999
## 11899	2165000	1230	1999
## 11982	3175000	7640	2007
## 11992	4311000	1670	1964
## 12212	1200000	820	1932
## 12255	1730000	1650	1980
## 12256	1730000	3260	1980
## 12392	1450000	2140	1987
## 12472	2350000	3150	1955
## 12487	2200000	5000	1962
## 12577	2700000	5030	2005
## 12582	3950000	7070	1974
## 12643	3850000	4690	2009
## 12686	148200	3720	2006
## 12759	1700000	4380	1990
## 12764	2988000	6990	2006
## 12816	2050000	5270	2015

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

```
sale_varsdf[sale_varsdf$large_residuals == TRUE, c("cooks", "leverage", "covratio")]
```

##	cooks	leverage	covratio
## 6	0.0004664609	2.893507e-04	0.9993949
## 25	0.0009753878	5.278467e-04	0.9994688
## 115	0.0020299462	5.845638e-04	0.9983898
## 178	0.0018244399	9.228660e-04	0.9997742
## 239	0.0002198987	1.479589e-04	0.9993415
## 246	0.0066006892	1.561306e-03	0.9988414
## 287	0.0010683643	4.772308e-04	0.9991448
## 295	0.0012422792	6.082382e-04	0.9994132
## 300	0.0064548391	1.591739e-03	0.9989924
## 341	0.0005442432	3.857083e-04	0.9996320
## 359	0.0014782483	6.146711e-04	0.9991662
## 385	0.0043135187	1.224730e-03	0.9989968
## 396	0.0016694478	5.905161e-04	0.9988469
## 475	0.0008135770	4.637368e-04	0.9994699
## 482	0.0021170740	1.260140e-03	1.0003199
## 508	0.0004022547	2.183994e-04	0.9991632

## 528	0.0017586547	9.402696e-04	0.9998662
## 661	0.0057964869	2.601807e-03	1.0012840
## 670	0.0031300034	4.972622e-04	0.9963319
## 679	0.0146362888	2.968097e-03	0.9997635
## 784	0.0011803649	6.373256e-04	0.9995755
## 811	0.0060910722	6.817328e-04	0.9946753
## 853	0.0003225509	1.950450e-04	0.9992714
## 877	0.0009214220	4.319784e-04	0.9991734
## 916	0.0072367610	2.192226e-03	1.0001222
## 1009	0.0004091340	2.609966e-04	0.9993977
## 1119	0.0037010280	1.235103e-03	0.9993744
## 1142	0.0027194222	1.022383e-03	0.9993965
## 1155	0.0013170704	4.634135e-04	0.9987091
## 1305	0.0021113566	1.320236e-03	1.0004367
## 1345	0.0003479257	2.599336e-04	0.9995568
## 1380	0.0003464408	2.009067e-04	0.9992279
## 1442	0.0014502269	6.768441e-04	0.9994119
## 1492	0.0006475096	4.204073e-04	0.9995764
## 1504	0.0079385980	2.305177e-03	1.0001359
## 1550	0.0015491030	7.182887e-04	0.9994436
## 1633	0.0003408684	1.975727e-04	0.9992239
## 1650	0.0237019887	3.888704e-03	0.9998782
## 1716	0.0013617758	4.526234e-04	0.9985820
## 1745	0.0020676432	7.101284e-04	0.9989076
## 1870	0.0122987518	5.588886e-03	1.0043155
## 1962	0.0031283954	6.832116e-04	0.9977157
## 1963	0.0003982272	1.948616e-04	0.9989986
## 1964	0.0017909124	3.951219e-04	0.9974597
## 1976	0.0017640939	2.339192e-04	0.9951983
## 1977	0.0013934335	1.674806e-04	0.9945890
## 1978	0.0014328789	1.767488e-04	0.9947468
## 1979	0.0013645416	1.647790e-04	0.9946134
## 1980	0.0013535941	1.628304e-04	0.9945893
## 1981	0.0020084343	2.793419e-04	0.9954890
## 1982	0.0016356662	2.109031e-04	0.9950260
## 2020	0.0012221030	8.367764e-04	1.0000491
## 2022	0.0029495101	8.578213e-04	0.9986875
## 2099	0.0010943702	5.673553e-04	0.9994517
## 2137	0.0007736426	2.304123e-04	0.9981156
## 2157	0.0009897771	3.861360e-04	0.9988267
## 2257	0.0003150914	2.045951e-04	0.9993605
## 2264	0.0038980620	8.708916e-04	0.9979757
## 2302	0.0026322104	1.229012e-03	0.9999658
## 2360	0.0007899378	3.477326e-04	0.9989921
## 2361	0.0030113026	5.568932e-04	0.9970108
## 2469	0.0004280295	1.613032e-04	0.9985386
## 2604	0.0002089868	1.424471e-04	0.9993493
## 2684	0.0013336264	1.688865e-04	0.9948856
## 2685	0.0015864972	2.125071e-04	0.9952298
## 2686	0.0024463729	3.886704e-04	0.9962233
## 2687	0.0014664406	1.894597e-04	0.9950153
## 2688	0.0012815502	1.591432e-04	0.9947669
## 2689	0.0022381170	3.433787e-04	0.9960219
## 2690	0.0023152329	3.603098e-04	0.9961032

```

## 2699 0.0011678370 6.853058e-04 0.9997270
## 2708 0.0018403998 4.829746e-04 0.9980520
## 2709 0.0021098330 6.231975e-04 0.9984894
## 2710 0.0019749571 5.489309e-04 0.9982666
## 2717 0.0037587786 7.228403e-04 0.9973217
## 2742 0.0006546637 2.806310e-04 0.9988822
## 2852 0.0585035861 7.417413e-03 1.0021976
## 2934 0.0013605077 4.526234e-04 0.9985839
## 2937 0.0013140035 4.526234e-04 0.9986557
## 3097 0.0034087423 6.913469e-04 0.9974783
## 3102 0.0012417631 4.821890e-04 0.9989145
## 3110 0.0007994239 3.874035e-04 0.9991773
## 3111 0.0005518095 3.464887e-04 0.9994657
## 3168 0.0063615130 3.373224e-04 0.9874295
## 3169 0.0063615130 3.373224e-04 0.9874295
## 3170 0.0049800930 2.754245e-04 0.9879069
## 3171 0.0041938513 2.357635e-04 0.9880707
## 3172 0.0043116463 2.414307e-04 0.9880275
## 3173 0.0041938513 2.357635e-04 0.9880707
## 3174 0.0043116463 2.414307e-04 0.9880275
## 3175 0.0054390063 2.947875e-04 0.9876701
## 3176 0.0040859024 2.326081e-04 0.9882221
## 3177 0.0038271713 2.180110e-04 0.9882149
## 3178 0.0038271713 2.180110e-04 0.9882149
## 3179 0.0047751411 2.635592e-04 0.9878698
## 3180 0.0038271713 2.180110e-04 0.9882149
## 3181 0.0047751411 2.635592e-04 0.9878698
## 3182 0.0063615130 3.373224e-04 0.9874295
## 3183 0.0063615130 3.373224e-04 0.9874295
## 3184 0.0063615130 3.373224e-04 0.9874295
## 3185 0.0063615130 3.373224e-04 0.9874295
## 3186 0.0063615130 3.373224e-04 0.9874295
## 3187 0.0063615130 3.373224e-04 0.9874295
## 3188 0.0063615130 3.373224e-04 0.9874295
## 3189 0.0054390063 2.947875e-04 0.9876701
## 3190 0.0043116463 2.414307e-04 0.9880275
## 3191 0.0054390063 2.947875e-04 0.9876701
## 3192 0.0041938513 2.357635e-04 0.9880707
## 3193 0.0043116463 2.414307e-04 0.9880275
## 3194 0.0043116463 2.414307e-04 0.9880275
## 3195 0.0038271713 2.180110e-04 0.9882149
## 3196 0.0038271713 2.180110e-04 0.9882149
## 3197 0.0040859024 2.326081e-04 0.9882221
## 3198 0.0063615130 3.373224e-04 0.9874295
## 3199 0.0063615130 3.373224e-04 0.9874295
## 3200 0.0063615130 3.373224e-04 0.9874295
## 3201 0.0063615130 3.373224e-04 0.9874295
## 3202 0.0063615130 3.373224e-04 0.9874295
## 3260 0.0005713790 4.231622e-04 0.9997120
## 3424 0.0008288589 3.301445e-04 0.9988075
## 3464 0.0053372173 2.947875e-04 0.9879098
## 3465 0.0048859524 2.754245e-04 0.9881441
## 3466 0.0037537572 2.180110e-04 0.9884487
## 3467 0.0037537572 2.180110e-04 0.9884487

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## 3468 0.0053372173 2.947875e-04 0.9879098
## 3469 0.0046849661 2.635592e-04 0.9881072
## 3470 0.0037537572 2.180110e-04 0.9884487
## 3471 0.0048859524 2.754245e-04 0.9881441
## 3472 0.0041139291 2.357635e-04 0.9883060
## 3473 0.0042296400 2.414307e-04 0.9882633
## 3474 0.0053372173 2.947875e-04 0.9879098
## 3475 0.0041139291 2.357635e-04 0.9883060
## 3476 0.0037880303 2.197074e-04 0.9884344
## 3477 0.0041139291 2.357635e-04 0.9883060
## 3478 0.0062437444 3.373224e-04 0.9876717
## 3479 0.0062437444 3.373224e-04 0.9876717
## 3480 0.0062437444 3.373224e-04 0.9876717
## 3481 0.0062437444 3.373224e-04 0.9876717
## 3482 0.0062437444 3.373224e-04 0.9876717
## 3483 0.0062437444 3.373224e-04 0.9876717
## 3484 0.0062437444 3.373224e-04 0.9876717
## 3485 0.0041297344 2.408396e-04 0.9885208
## 3486 0.0046849661 2.635592e-04 0.9881072
## 3487 0.0039371732 2.291360e-04 0.9884845
## 3488 0.0042296400 2.414307e-04 0.9882633
## 3489 0.0041139291 2.357635e-04 0.9883060
## 3490 0.0042296400 2.414307e-04 0.9882633
## 3491 0.0037537572 2.180110e-04 0.9884487
## 3492 0.0048859524 2.754245e-04 0.9881441
## 3493 0.0037537572 2.180110e-04 0.9884487
## 3494 0.0053372173 2.947875e-04 0.9879098
## 3495 0.0062437444 3.373224e-04 0.9876717
## 3496 0.0062437444 3.373224e-04 0.9876717
## 3497 0.0062437444 3.373224e-04 0.9876717
## 3523 0.0031516419 8.461259e-04 0.9984756
## 3837 0.0009035567 4.093318e-04 0.9990987
## 3918 0.0023316105 1.235103e-03 1.0001495
## 3919 0.0017661073 1.022383e-03 1.0000484
## 4055 0.0009190774 3.045701e-04 0.9984275
## 4056 0.0076748548 9.372600e-04 0.9954514
## 4285 0.0005040196 2.917156e-04 0.9993164
## 4435 0.0006298517 3.527651e-04 0.9993371
## 4648 0.0249625344 9.532013e-04 0.9829702
## 4649 0.0412679063 1.029374e-03 0.9734661
## 4671 0.0039974176 2.311265e-03 1.0013404
## 4695 0.0021082688 2.486567e-04 0.9945596
## 4696 0.0023221566 6.979676e-04 0.9986051
## 4740 0.0003376386 2.147132e-04 0.9993479
## 4750 0.0030377724 2.192226e-03 1.0014613
## 4821 0.0009996267 3.569086e-04 0.9986314
## 4834 0.0025288899 8.578213e-04 0.9990301
## 4840 0.0013835219 4.634135e-04 0.9986089
## 4934 0.0018549558 8.318031e-04 0.9995060
## 5083 0.0050526138 1.436153e-03 0.9992115
## 5491 0.0002523170 1.635798e-04 0.9993177
## 5494 0.0002484208 1.647050e-04 0.9993428
## 5495 0.0002523170 1.635798e-04 0.9993177
## 5496 0.0002523170 1.635798e-04 0.9993177

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## 5497 0.0002531760 1.634073e-04 0.9993128
## 5549 0.0004600685 3.168289e-04 0.9995343
## 5935 0.0024867718 8.422314e-04 0.9990113
## 6055 0.0003145941 2.149613e-04 0.9994244
## 6230 0.0020591500 1.897492e-04 0.9928470
## 6231 0.0023651015 2.270410e-04 0.9931872
## 6232 0.0024525684 2.426356e-04 0.9934181
## 6233 0.0020979429 1.884034e-04 0.9926483
## 6234 0.0020695093 1.913460e-04 0.9928740
## 6235 0.0020591500 1.897492e-04 0.9928470
## 6236 0.0020695093 1.913460e-04 0.9928740
## 6237 0.0021498317 1.986234e-04 0.9928756
## 6238 0.0020913876 1.850418e-04 0.9925288
## 6239 0.0023651015 2.270410e-04 0.9931872
## 6429 0.0061300865 1.874854e-04 0.9777108
## 6430 0.0059202411 1.667110e-04 0.9757505
## 6431 0.0059797139 1.736889e-04 0.9765041
## 6432 0.0058428794 1.739568e-04 0.9770827
## 6433 0.0054544428 1.648867e-04 0.9774238
## 6434 0.0059907580 1.816395e-04 0.9775083
## 6435 0.0069990643 2.210568e-04 0.9784576
## 6436 0.0059120325 1.726999e-04 0.9766372
## 6437 0.0056564746 1.607759e-04 0.9759715
## 6438 0.0063120189 1.778323e-04 0.9757739
## 6439 0.0056603235 1.707076e-04 0.9773758
## 6440 0.0063585188 1.965568e-04 0.9779589
## 6441 0.0058380478 1.729858e-04 0.9769711
## 6442 0.0058955217 1.772123e-04 0.9773064
## 6443 0.0050977569 1.723986e-04 0.9798524
## 6444 0.0052311711 1.828160e-04 0.9805223
## 6445 0.0059981754 1.953223e-04 0.9790890
## 6446 0.0053671634 1.830989e-04 0.9800403
## 6447 0.0056716563 2.034328e-04 0.9810504
## 6448 0.0051797909 1.693365e-04 0.9791460
## 6449 0.0059801345 2.172168e-04 0.9813053
## 6450 0.0070060323 2.627101e-04 0.9819468
## 6451 0.0056326441 1.833206e-04 0.9790655
## 6452 0.0053652371 1.834080e-04 0.9800820
## 6453 0.0062745117 2.290056e-04 0.9814082
## 6454 0.0053414175 1.882225e-04 0.9806909
## 6455 0.0051374917 1.725104e-04 0.9797068
## 6456 0.0053106364 1.756395e-04 0.9793957
## 6457 0.0053903273 1.905389e-04 0.9807542
## 6512 0.0034210414 6.510037e-04 0.9972114
## 6527 0.0011519050 5.869196e-04 0.9994477
## 6739 0.0005422379 2.629729e-04 0.9990539
## 6796 0.0004259225 2.974720e-04 0.9995291
## 6821 0.0016510477 9.316549e-04 0.9999262
## 6931 0.0004079721 2.499580e-04 0.9993414
## 6938 0.0013332318 2.843728e-04 0.9972400
## 6939 0.0017438684 4.410976e-04 0.9979102
## 6940 0.0010991039 2.017637e-04 0.9966273
## 6941 0.0011010038 2.025925e-04 0.9966371
## 6942 0.0015919676 3.791323e-04 0.9976767

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## 6943 0.0015852546 3.765171e-04 0.9976661
## 6944 0.0018719324 4.981012e-04 0.9981037
## 6945 0.0013332318 2.843728e-04 0.9972400
## 6946 0.0011010038 2.025925e-04 0.9966371
## 6947 0.0010844651 1.944384e-04 0.9965292
## 6948 0.0010802487 1.914325e-04 0.9964805
## 7039 0.0017636293 9.841865e-04 0.9999650
## 7147 0.0016989320 7.235079e-04 0.9993149
## 7167 0.0080037666 6.011161e-04 0.9915447
## 7210 0.0034800889 5.546969e-04 0.9964038
## 7211 0.0018864790 2.561550e-04 0.9953439
## 7446 0.0036456431 1.876792e-04 0.9868871
## 7447 0.0038433663 1.946625e-04 0.9866729
## 7448 0.0036392849 1.874697e-04 0.9868953
## 7449 0.0038433663 1.946625e-04 0.9866729
## 7450 0.0038522383 1.949902e-04 0.9866647
## 7451 0.0038612057 1.953223e-04 0.9866565
## 7452 0.0038433663 1.946625e-04 0.9866729
## 7453 0.0034644328 1.775582e-04 0.9868169
## 7454 0.0034519273 1.771395e-04 0.9868334
## 7455 0.0036098063 1.827092e-04 0.9866517
## 7456 0.0038433663 1.946625e-04 0.9866729
## 7457 0.0036586326 1.881113e-04 0.9868706
## 7458 0.0038612057 1.953223e-04 0.9866565
## 7459 0.0039865399 2.000377e-04 0.9865499
## 7460 0.0038522383 1.949902e-04 0.9866647
## 7461 0.0039266629 1.977694e-04 0.9865991
## 7462 0.0039561638 1.988839e-04 0.9865745
## 7463 0.0039266629 1.977694e-04 0.9865991
## 7507 0.0014626494 6.373256e-04 0.9992658
## 7649 0.0005467171 3.750522e-04 0.9995887
## 7650 0.0005467171 3.750522e-04 0.9995887
## 7791 0.0011445250 8.396526e-04 1.0001201
## 7871 0.0043052636 1.418340e-03 0.9995310
## 8119 0.0013458758 9.316549e-04 1.0001553
## 8154 0.0004468693 2.899718e-04 0.9994452
## 8232 0.0014741764 6.373256e-04 0.9992531
## 8262 0.0445055400 1.048109e-02 1.0078588
## 8320 0.0023704331 9.403814e-04 0.9994116
## 8377 0.0263960969 3.400356e-03 0.9982227
## 8457 0.0003857287 1.948616e-04 0.9990434
## 8458 0.0017530634 3.951219e-04 0.9975266
## 8535 0.0006474578 3.931427e-04 0.9994745
## 8541 0.0003429674 1.975727e-04 0.9992165
## 8698 0.0003905479 2.127067e-04 0.9991616
## 8710 0.0041172837 5.668439e-04 0.9957255
## 8763 0.0006975138 3.391819e-04 0.9991340
## 8887 0.0128047407 1.244190e-03 0.9942935
## 8911 0.0021982138 1.216459e-03 1.0001873
## 8946 0.0021982138 1.216459e-03 1.0001873
## 9215 0.0012471199 4.801058e-04 0.9988968
## 9293 0.0018004712 9.402696e-04 0.9998351
## 9420 0.0008278908 5.117880e-04 0.9996137
## 9453 0.0007098136 4.526234e-04 0.9995890

```

```
## 9528 0.0100392020 1.772567e-03 0.9980507
## 9546 0.0008646845 4.093318e-04 0.9991651
## 9722 0.0005923976 2.147132e-04 0.9985183
## 10125 0.0026517491 1.522336e-03 1.0005396
## 10318 0.0017609131 6.468996e-04 0.9989768
## 10371 0.0010176808 5.615556e-04 0.9995274
## 10418 0.0006858285 4.735117e-04 0.9996937
## 10623 0.0006563793 4.477275e-04 0.9996556
## 10707 0.0001445453 9.054752e-05 0.9992070
## 10723 0.0007332636 4.821890e-04 0.9996519
## 10741 0.0004678762 3.464935e-04 0.9996352
## 10787 0.0052823602 1.216153e-03 0.9984143
## 10844 0.0031108380 1.582041e-03 1.0004426
## 10958 0.0005425057 2.535700e-04 0.9989903
## 10995 0.0055155406 3.612158e-03 1.0027911
## 11165 0.0006900799 2.933313e-04 0.9988811
## 11289 0.0011758876 1.908893e-04 0.9961190
## 11413 0.0038167484 6.738844e-04 0.9969489
## 11558 0.0059203676 6.219845e-04 0.9942086
## 11586 0.0034271108 9.967534e-04 0.9988267
## 11728 0.0004760078 1.156990e-04 0.9974723
## 11758 0.0010922881 5.615556e-04 0.9994345
## 11772 0.0211815375 1.439821e-03 0.9914144
## 11822 0.0028122280 3.096546e-04 0.9942002
## 11898 0.0007945653 1.827581e-04 0.9973763
## 11899 0.0020091839 2.625214e-04 0.9951489
## 11982 0.0147427616 2.192226e-03 0.9977314
## 11992 0.0121394901 3.095031e-04 0.9733415
## 12212 0.0028335562 1.089707e-03 0.9995056
## 12255 0.0006600813 1.602514e-04 0.9975135
## 12256 0.0005042011 2.009995e-04 0.9986797
## 12392 0.0001886739 9.473346e-05 0.9989348
## 12472 0.0044534605 5.997623e-04 0.9956455
## 12487 0.0042234880 1.121244e-03 0.9987220
## 12577 0.0037179263 5.703393e-04 0.9962485
## 12582 0.0386471549 2.245096e-03 0.9904841
## 12643 0.0090749332 4.506111e-04 0.9866549
## 12686 0.0002951996 2.013046e-04 0.9994087
## 12759 0.0005506545 3.940090e-04 0.9996497
## 12764 0.0102988857 1.682586e-03 0.9976412
## 12816 0.0013304277 6.853058e-04 0.9995611
```

There are no problematic Cook's values since there are none greater than 1. The leverage values all lie between 0 and 1 so there are no problematic indicators there either. The covariance ratios also seem to be more or less within the approximate boundaries (between 1.0007 and 0.9993), however, there are a few that fall around the 0.98 range.

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

```
durbinWatsonTest(sale_vars)
```



```
## lag Autocorrelation D-W Statistic p-value
## 1 0.7195885 0.5608136 0
## Alternative hypothesis: rho != 0
```

The p-value is 0 which suggests the variables do significantly impact the outcome.

m. Perform the necessary calculations to assess the assumption of no multi-collinearity and state if the condition is met or not.

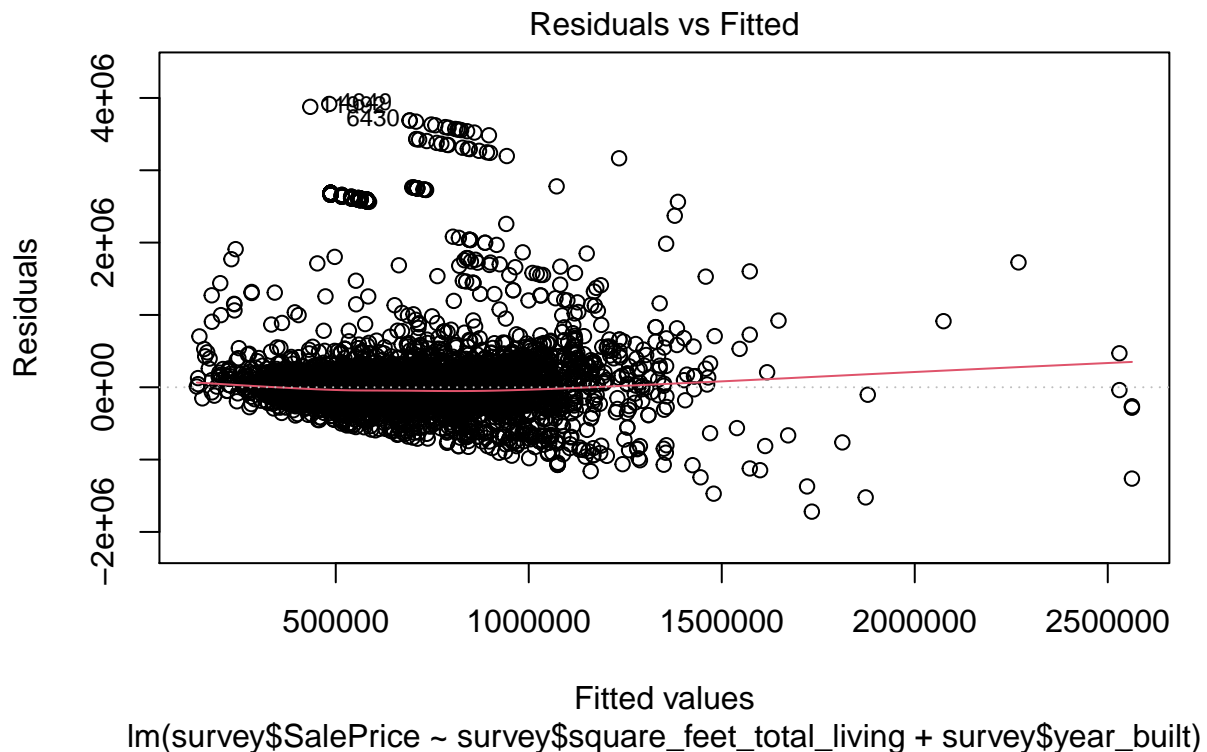
```
vif(sale_vars)
```

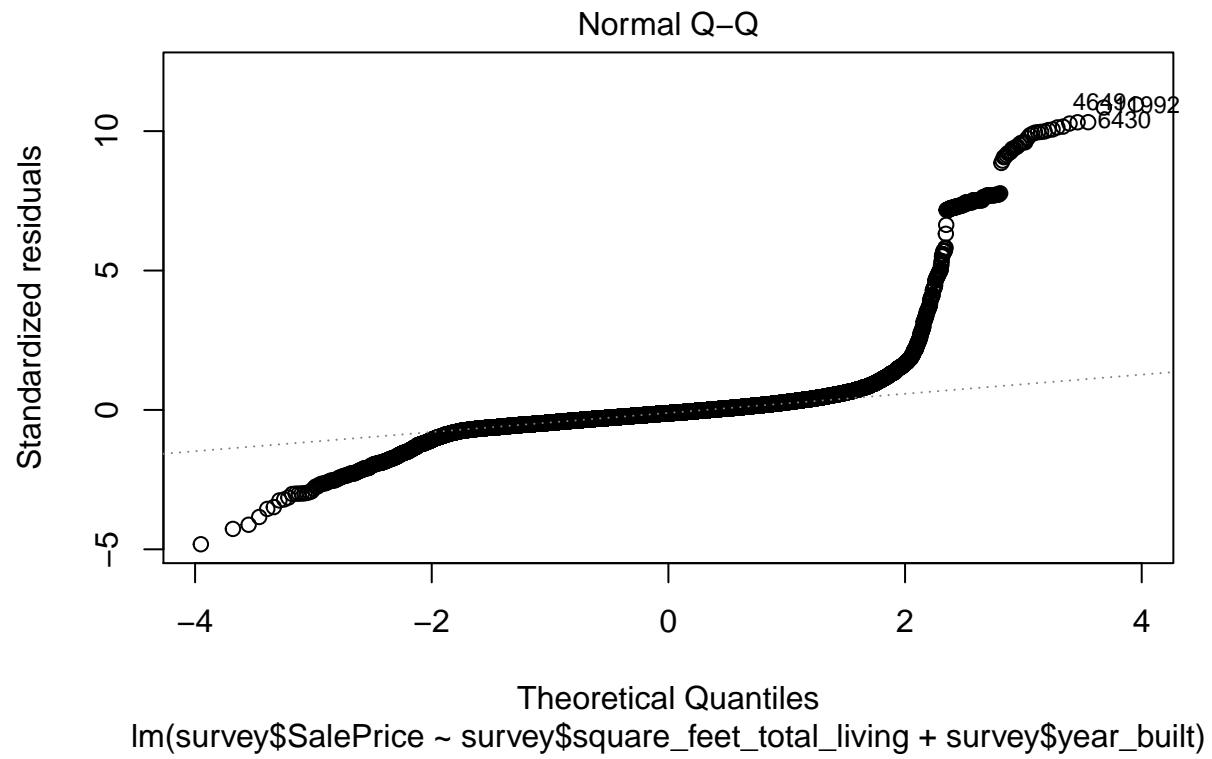
```
## survey$square_feet_total_living survey$year_built
## 1.103628 1.103628
```

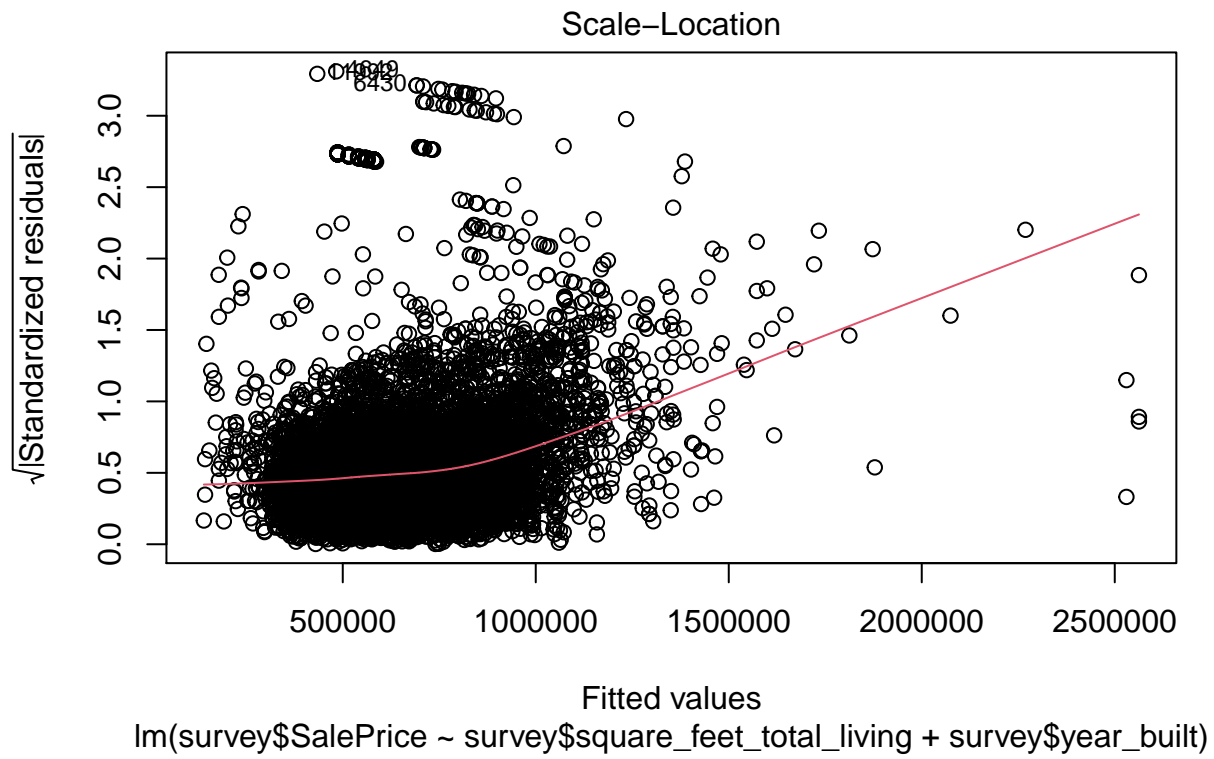
Both VIF figures are very close to 1 so we can conclude that there is no collinearity in the data.

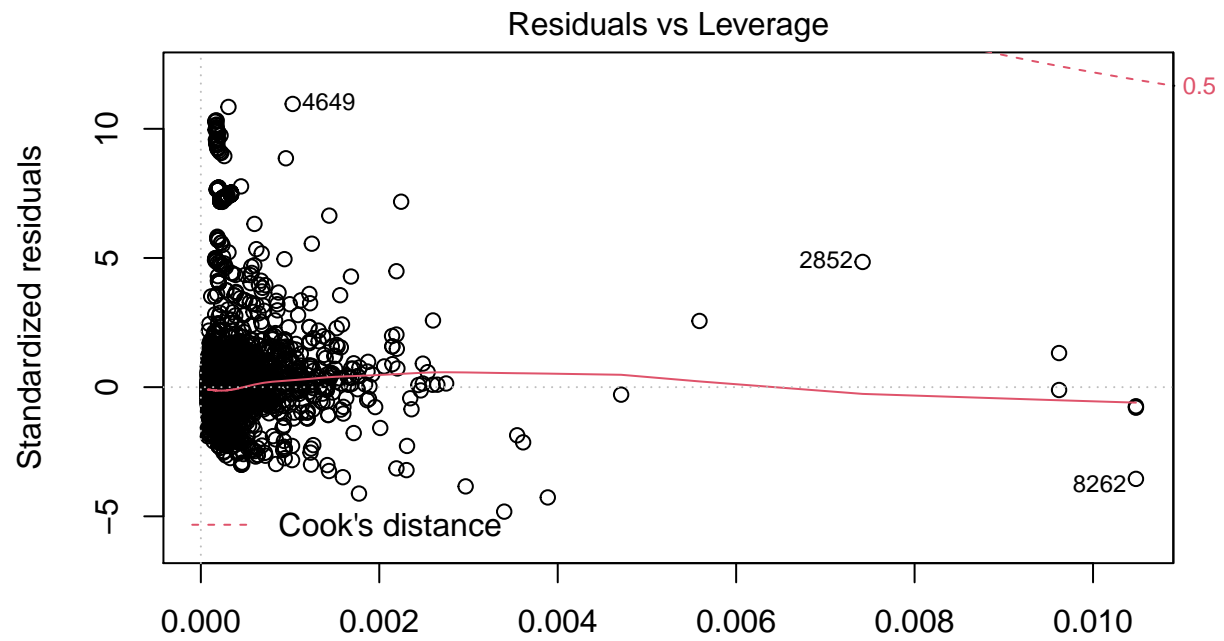
n. Visually check the assumptions related to the residuals using the plot() and hist() functions. Summarize what each graph is informing you of and if any anomalies are present.

```
plot(sale_vars)
```





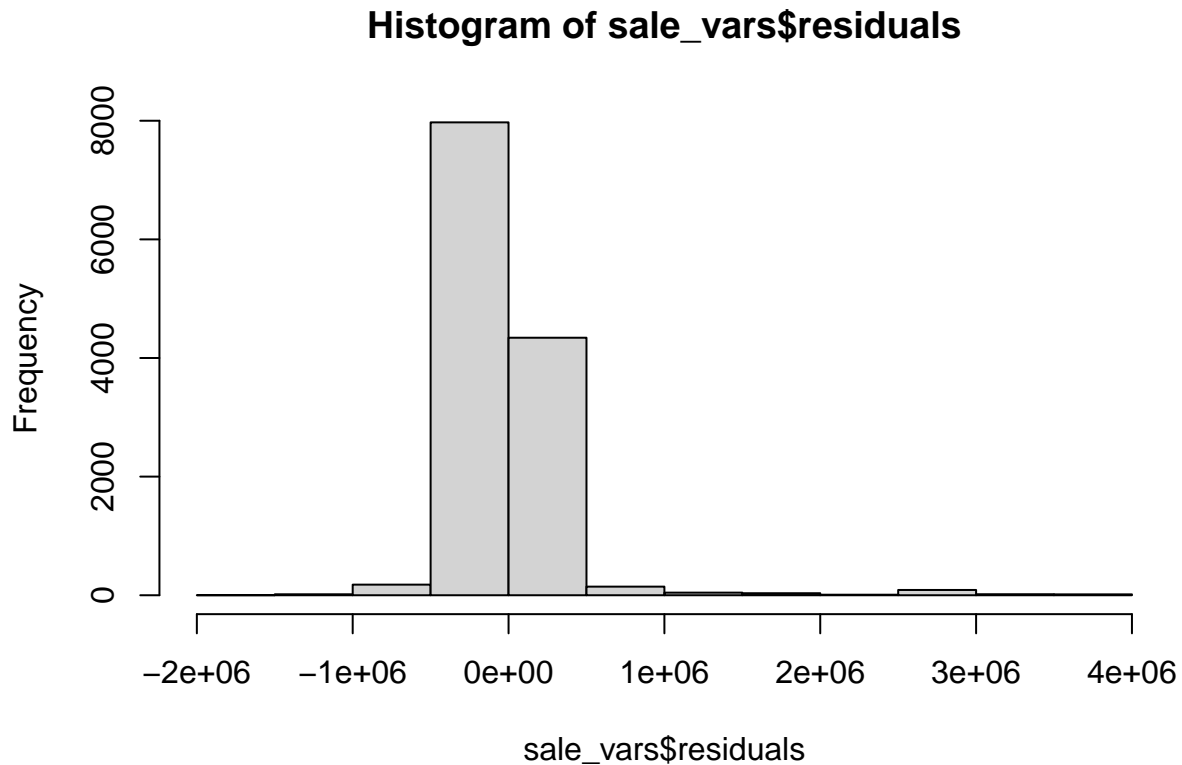




Leverage

$\text{lm}(\text{survey}\$SalePrice \sim \text{survey}\$square\_feet\_total\_living + \text{survey}\$year\_built)$

```
hist(sale_vars$residuals)
```



By looking at the graph, it seems like the data may not be linear. There seems to be some outliers far out from the majority of the data in both directions. The normal Q-Q graph in particular shows a dip to the right in the mid-range of the standardized residuals.

**o. Overall, is this regression model unbiased? If an unbiased regression model, what does this tell us about the sample vs. the entire population model?**

With the calculations done with the data, it seems that the data is largely unbiased. There may be some outliers but they may not be statistically significant enough to claim that the observations would be a poor representation of the entire population.