

Final

David Moste

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Problem 1

I started by creating all four sets: X, Y, x, and y. I then used R to calculate all the listed probabilities.

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.0 --

## v ggplot2 3.3.2      v purrr 0.3.4
## v tibble 3.0.3       v dplyr 1.0.2
## v tidyr 1.1.2        v stringr 1.4.0
## v readr 1.3.1        v forcats 0.5.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

set.seed(12345)
N <- 10
mu <- (N+1)/2
sigma <- (N+1)/2
X <- runif(10000,1,N)
Y <- rnorm(10000,mean = mu,sd = sigma)
df <- data.frame(X,Y)

x <- median(X)
y <- summary(Y)[2]

df_Xy <- df %>% filter(X > y)
df_Yy <- df %>% filter(Y > y)

# Part a
df_a <- df_Xy %>% filter(X > x)
prob_a <- nrow(df_a)/nrow(df_Xy)
prob_a

## [1] 0.5512679
```

```
# Part b
df_b <- df_Yy %>% filter(X > x)
prob_b <- nrow(df_b)/nrow(df)
prob_b
```

```
## [1] 0.3808
```

```
# Part c
df_c <- df_Xy %>% filter(X < x)
prob_c <- nrow(df_c)/nrow(df_Xy)
prob_c
```

```
## [1] 0.4487321
```

$P_A = 0.55$

$P_B = 0.38$

$P_C = 0.45$

The next step was to create the marginal and joint probability table.

```
# Probability table
XY <- df %>% filter(X > x, Y > y) %>% nrow()
Xy <- df %>% filter(X > x, Y < y) %>% nrow()
Xx_total <- XY/nrow(df) + Xy/nrow(df)

xY <- df %>% filter(X < x, Y > y) %>% nrow()
xy <- df %>% filter(X < x, Y < y) %>% nrow()
xX_total <- xY/nrow(df) + xy/nrow(df)

Yy_total <- XY/nrow(df) + xY/nrow(df)
yY_total <- Xy/nrow(df) + xy/nrow(df)
total <- Xx_total + xX_total

table <- data.frame(rbind(XY/nrow(df),Xy/nrow(df),XY/nrow(df) + Xy/nrow(df)),
                    rbind(xY/nrow(df),xy/nrow(df),xY/nrow(df) + xy/nrow(df)),
                    rbind(Yy_total,yY_total,total))
names(table) <- c("X > x","X < x","Total")
row.names(table) <- c("Y > y","Y < y","Total")
table
```

```
##           X > x  X < x Total
## Y > y 0.3808 0.3692  0.75
## Y < y 0.1192 0.1308  0.25
## Total 0.5000 0.5000  1.00
```

As can be seen, it is clear that the two probabilities are equal.

$$P(X > x, Y > y) = P(X > x) P(Y > y)$$

The next step is to check for independence. Fisher's Test works for small datasets while Chi Squared is more appropriate for larger datasets. Therefore I would choose Chi Squared for this problem.

```
# Check independence
count_table <- matrix(c(XY,Xy,xY,xy),nrow = 2, ncol = 2, byrow = FALSE)
fisher.test(count_table)
```

```
##
## Fisher's Exact Test for Count Data
##
## data: count_table
## p-value = 0.007904
## alternative hypothesis: true odds ratio is not equal to 1
## 95 percent confidence interval:
## 1.032680 1.240419
## sample estimates:
## odds ratio
## 1.131777
```

```
chisq.test(count_table)
```

```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: count_table
## X-squared = 7.0533, df = 1, p-value = 0.007912
```

Problem 2

Descriptive and Inferential Statistics

```
train <- read.csv('https://raw.githubusercontent.com/dmoste/DATA605/master/train.csv')
test <- read.csv('https://raw.githubusercontent.com/dmoste/DATA605/master/test.csv')

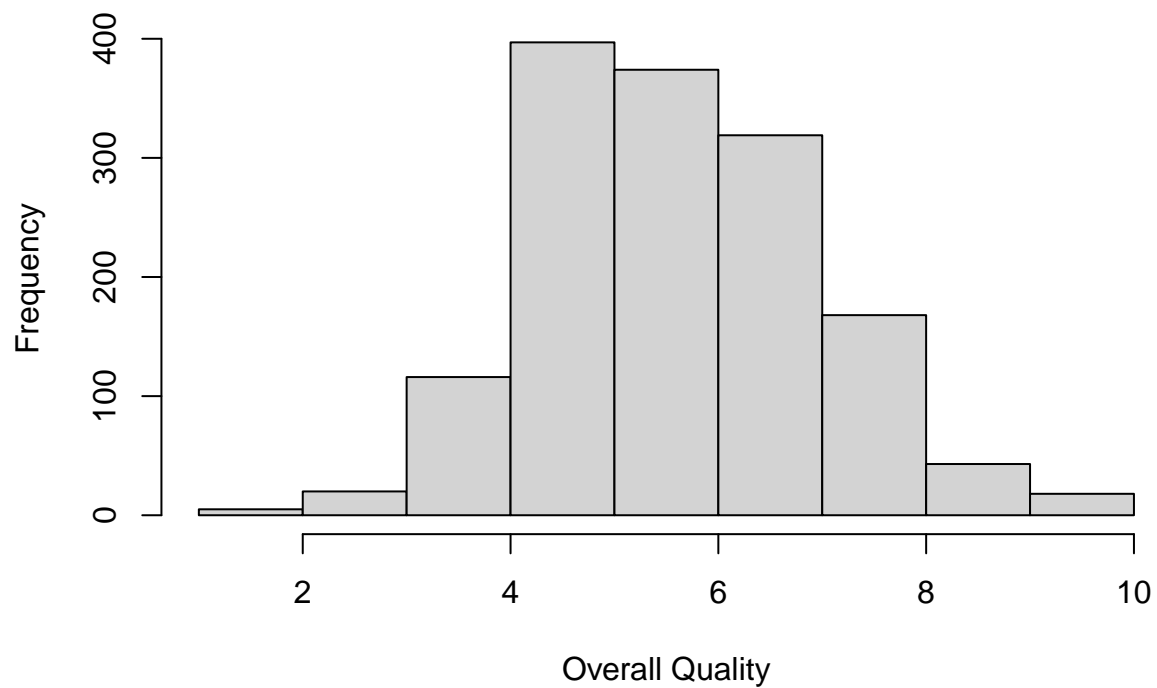
train[is.na(train)] = 0

# Summary and histogram of Overall Quality
summary(train$OverallQual)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.000   5.000   6.000   6.099   7.000   10.000
```

```
hist(train$OverallQual,
     xlab = "Overall Quality",
     main = "Histogram of Overall Quality")
```

Histogram of Overall Quality

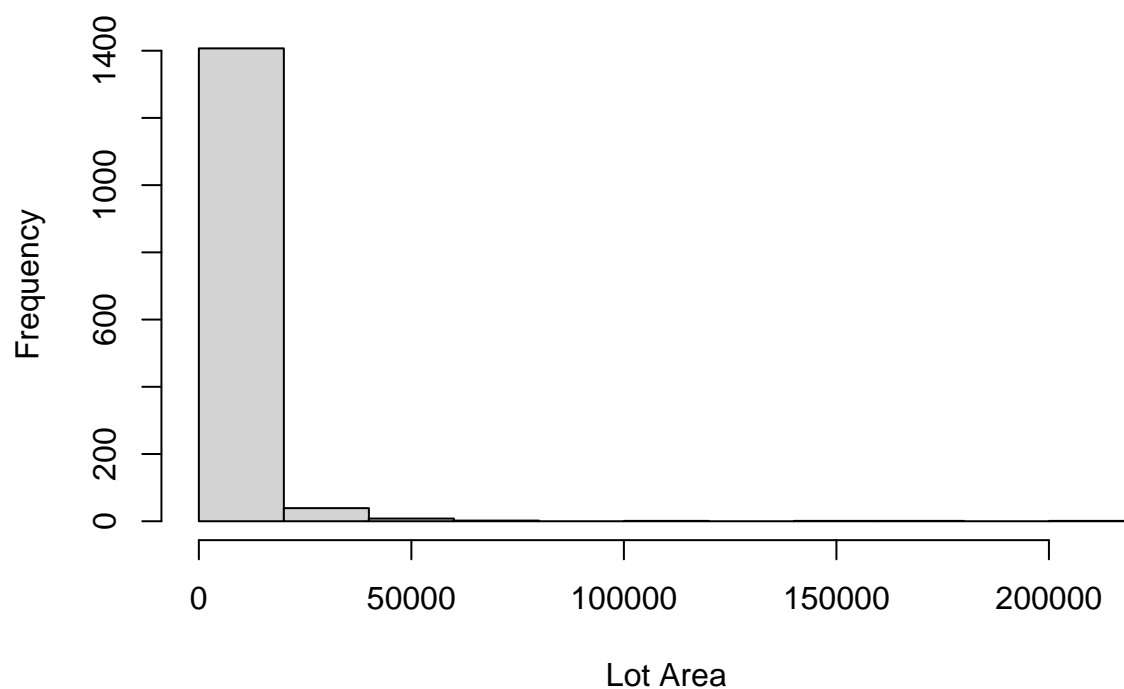


```
# Summary and histogram of Lot Area  
summary(train$LotArea)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   
##      1300   7554   9478   10517   11602   215245
```

```
hist(train$LotArea,  
      xlab = "Lot Area",  
      main = "Histogram of Lot Area")
```

Histogram of Lot Area

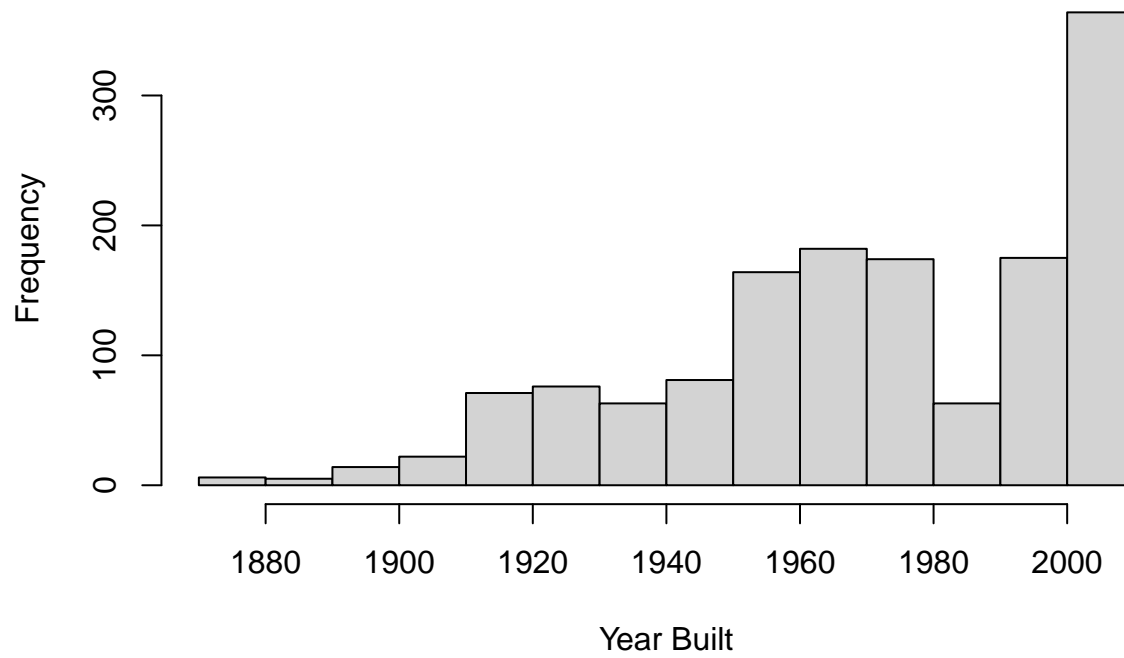


```
# Summary and histogram of Year Built  
summary(train$YearBuilt)
```

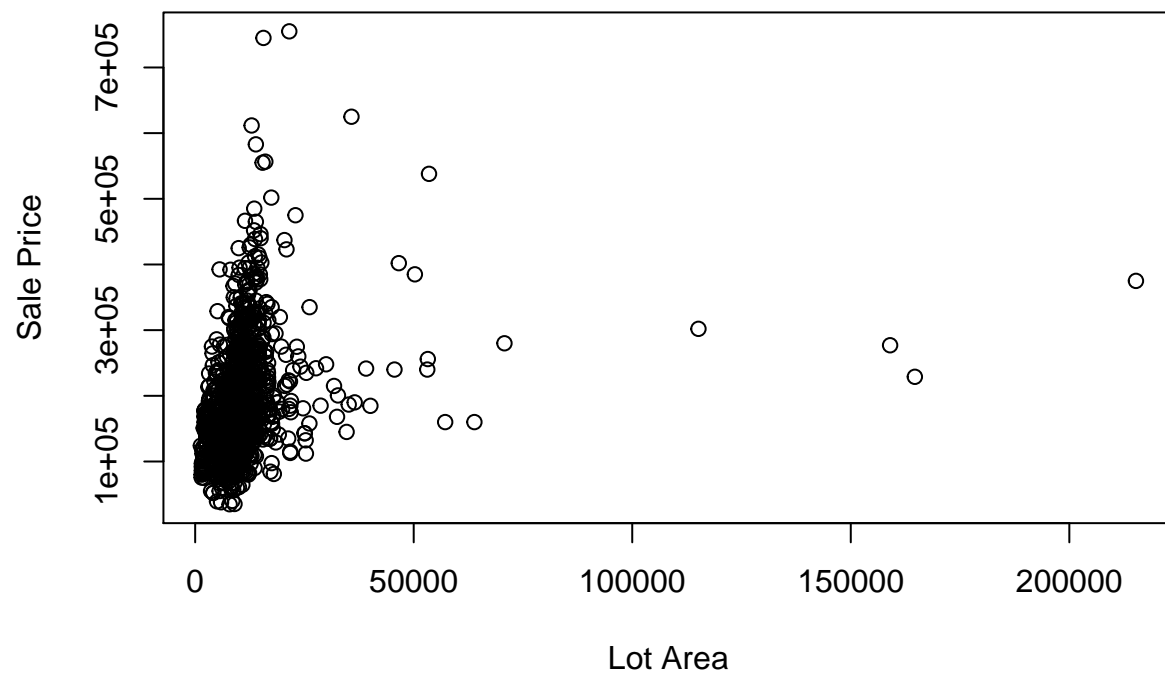
```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   
##      1872   1954   1973   1971   2000   2010
```

```
hist(train$YearBuilt,  
      xlab = "Year Built",  
      main = "Histogram of Year Built")
```

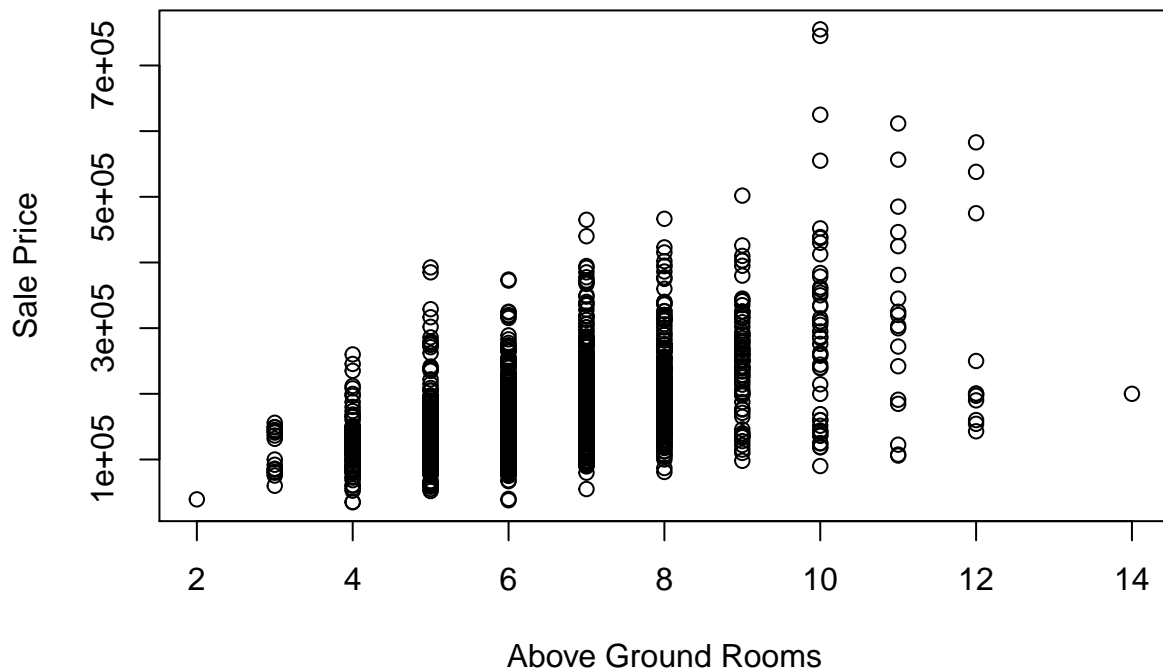
Histogram of Year Built



```
# Plot Lot Area vs Sale Price  
plot(train$LotArea,  
      train$SalePrice,  
      xlab = "Lot Area",  
      ylab = "Sale Price")
```



```
# Plot Total above ground rooms vs Sale Price
plot(train$TotRmsAbvGrd,
      train$SalePrice,
      xlab = "Above Ground Rooms",
      ylab = "Sale Price")
```



```
# Remove outliers
train <- train %>% filter(LotArea < 100000)

# Create a correlation matrix for Lot Area, Year Built, and Sale Price
correlation <- cor(train %>% dplyr::select(LotArea, YearBuilt, SalePrice) %>% as.matrix())
correlation

##           LotArea  YearBuilt SalePrice
## LotArea   1.00000000  0.04230918  0.3544944
## YearBuilt  0.04230918  1.00000000  0.5255868
## SalePrice  0.35449443  0.52558678  1.0000000

# Run a correlation test with 80% confidence interval
cor.test(train$LotArea, train$SalePrice, conf.level = 0.8)

##
## Pearson's product-moment correlation
##
## data:  train$LotArea and train$SalePrice
## t = 14.456, df = 1454, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 80 percent confidence interval:
##  0.3247557 0.3835329
## sample estimates:
##      cor
## 0.3544944
```



```
cor.test(train$YearBuilt, train$SalePrice, conf.level = 0.8)
```

```
##
## Pearson's product-moment correlation
##
## data: train$YearBuilt and train$SalePrice
## t = 23.558, df = 1454, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 80 percent confidence interval:
## 0.5008255 0.5494885
## sample estimates:
## cor
## 0.5255868
```

The correlation between Lot Area and Sale Price is 0.35, which indicates they are weakly correlated at best. The correlation between Year Built and Sale Price is 0.52, which is another weak correlation. The p-value is less than 0.05 though for both, which suggests the correlation is real.

Linear Algebra and Correlation

```
# Invert the correlation matrix
precision = solve(correlation)

# Multiply the correlation and precision matrices
round(correlation %*% precision)
```

```
##           LotArea YearBuilt SalePrice
## LotArea      1         0         0
## YearBuilt     0         1         0
## SalePrice     0         0         1
```

```
round(precision %*% correlation)
```

```
##           LotArea YearBuilt SalePrice
## LotArea      1         0         0
## YearBuilt     0         1         0
## SalePrice     0         0         1
```

```
# Conduct LU decomposition
library(matrixcalc)
```

```
## Warning: package 'matrixcalc' was built under R version 4.0.3
```

```
decomp <- matrixcalc::lu.decomposition(correlation)
decomp
```

```
## $L
##           [,1]      [,2] [,3]
```

```
## [1,] 1.00000000 0.000000 0
## [2,] 0.04230918 1.000000 0
## [3,] 0.35449443 0.511504 1
##
## $U
##      [,1]      [,2]      [,3]
## [1,]    1 0.04230918 0.3544944
## [2,]    0 0.99820993 0.5105884
## [3,]    0 0.00000000 0.6131657
```

Calculus-Based Probability & Statistics

```
library(MASS)
```

```
## Warning: package 'MASS' was built under R version 4.0.3
```

```
##
## Attaching package: 'MASS'
```

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

```
# Fit Exponential Probability Density
ep = MASS::fitdistr(train$LotArea, "exponential")
ep
```

```
##      rate
## 9.904415e-05
## (2.595662e-06)
```

```
# Find the optimal lambda
lambda = ep$estimate
lambda
```

```
##      rate
## 9.904415e-05
```

```
# Get 1000 values
samples = rexp(1000, lambda)
samples
```

```
##      [1] 12759.353408 6758.703443 1646.158875 14978.258847 4125.725679
##      [6] 3972.972632 19076.061572 2597.712978 14967.294877 14016.218359
##     [11] 920.934735 2680.105893 5310.812064 8079.675692 11522.474613
##     [16] 11538.935490 2144.935351 1592.732518 13517.936092 40556.932542
##     [21] 8722.787067 23959.528560 2228.853826 14166.285138 1043.722041
##     [26] 9975.405026 12082.506550 9471.273655 36416.087021 3684.155675
##     [31] 1955.995685 30477.861548 20337.597183 3186.218267 20151.545872
```

##	[36]	3862.866029	4148.178165	3170.693961	854.263960	25973.790525
##	[41]	33215.740557	6435.568467	1268.136506	21244.180496	3506.718879
##	[46]	10517.833929	23917.661910	20544.206145	4320.607816	15634.186493
##	[51]	21595.271765	17636.750649	4012.054044	2656.947318	19024.572725
##	[56]	2315.103519	2266.475401	9248.438920	14811.978680	10338.222691
##	[61]	7256.065942	24665.272543	17787.576267	7653.117254	1025.728318
##	[66]	4636.119880	2730.942597	5159.999017	2486.115937	3305.584220
##	[71]	1554.227525	4242.922685	2006.430690	5128.508375	29286.313772
##	[76]	669.389030	11192.717587	18585.455663	505.045197	21641.270356
##	[81]	15726.146180	674.594378	790.706901	765.385516	2240.183303
##	[86]	5624.937003	317.842570	12740.188440	11702.790121	31830.531486
##	[91]	10212.736952	28385.222982	1234.941766	13969.032500	4408.898826
##	[96]	9436.529802	21843.985821	10434.139435	5800.810703	9046.358672
##	[101]	1710.525182	5966.001130	20110.185241	41.587773	7709.612614
##	[106]	21375.947748	2148.600290	20482.141794	1029.798217	779.470567
##	[111]	16266.434663	13966.209077	12936.387347	767.215722	6809.852777
##	[116]	5269.500506	2980.571350	6515.954087	6920.203856	7347.088553
##	[121]	5361.034571	6034.129449	1231.155835	17982.337561	3839.503910
##	[126]	832.605173	17902.241234	4607.888469	34588.925162	24282.721304
##	[131]	30635.238730	3846.878728	76.164533	4003.398989	1883.077241
##	[136]	2003.089191	72717.001619	14456.087671	3726.302270	4421.688987
##	[141]	488.420939	8233.621744	4349.327389	3995.937408	8782.380507
##	[146]	3392.573755	14473.475721	7075.375824	2212.433047	1017.627837
##	[151]	13495.780204	1650.628242	33946.433913	2234.193764	21669.244477
##	[156]	9051.528384	1183.831393	4886.524305	27250.576675	4715.335810
##	[161]	9182.750055	3972.403302	2824.462743	12650.070319	8733.167108
##	[166]	1107.402786	6640.017136	8740.675327	17456.928835	19887.623402
##	[171]	1056.588745	3387.398805	21989.390478	4300.540984	9606.933931
##	[176]	2083.854043	8179.737860	7694.539605	16094.486431	4152.863113
##	[181]	4852.189899	4441.958216	4661.281292	33006.888784	4957.485248
##	[186]	1528.799629	791.319055	5020.780851	36.087572	8718.373760
##	[191]	11497.603228	12334.624432	20361.657096	12189.876059	841.612223
##	[196]	1880.603489	25010.466711	14696.607560	3333.462496	10833.266722
##	[201]	2660.485721	2073.717864	28869.849284	1816.209693	4616.066930
##	[206]	25782.775616	2439.775720	5164.194678	22944.772994	2741.843063
##	[211]	2490.977094	14602.894970	11073.511436	12248.674383	7848.279222
##	[216]	1719.334834	13011.007157	15201.494783	3711.463130	9961.034742
##	[221]	715.659455	5203.420198	13077.511824	2653.915969	77.237987
##	[226]	4087.488920	4524.220688	1300.886121	28618.648905	42612.728791
##	[231]	16640.285622	2208.775004	3453.553777	2938.458950	955.370639
##	[236]	3220.677287	34719.903048	10220.668791	6543.585215	12746.226583
##	[241]	6960.042285	10209.163591	6838.752977	2713.817831	6681.737135
##	[246]	31686.565128	6043.739277	2049.190092	465.035219	2225.499654
##	[251]	11770.824109	2552.104860	10589.996434	54.398733	1877.614905
##	[256]	4462.848034	2709.700045	10009.304067	7227.212387	14386.197844
##	[261]	103.089174	5869.357099	6873.651672	18606.600825	10729.253641
##	[266]	2221.760285	6577.071206	2778.509689	29206.282889	11333.419751
##	[271]	14823.799587	25941.275489	1744.361414	18814.791587	4756.039278
##	[276]	10723.743113	5849.094270	16869.421191	19466.617393	5947.830033
##	[281]	16677.841516	11617.831945	28514.718554	221.425539	8463.517170
##	[286]	262.429168	6342.918978	13974.125448	11095.223350	16820.157203
##	[291]	5170.241707	7857.933534	1898.726611	8506.388756	19149.505580
##	[296]	11494.700677	3696.098611	19830.463142	715.424479	5049.832540
##	[301]	33714.331646	22927.100910	3387.857923	4764.888591	186.585200

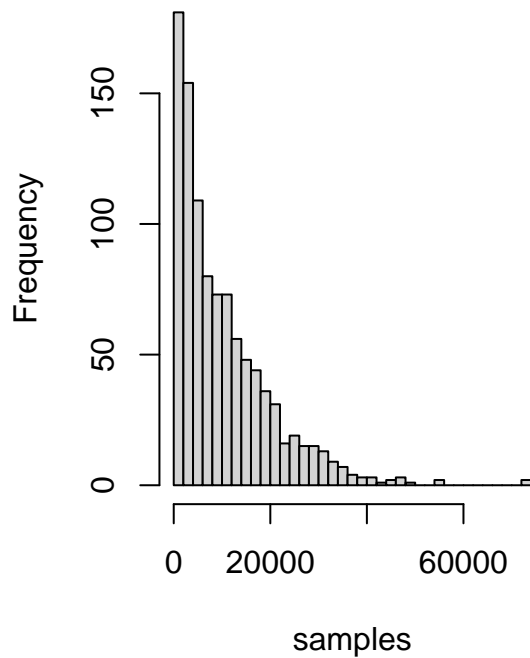
##	[306]	27949.671309	807.183373	784.677409	3507.483751	37642.631091
##	[311]	2048.173861	12117.436237	10674.458804	2901.917274	19674.273466
##	[316]	19768.534558	7194.941397	7571.269426	1583.961861	12304.356122
##	[321]	4848.450533	10814.814764	13718.570444	10753.812954	11474.882401
##	[326]	8743.813285	13658.346741	1367.188850	9373.933075	9115.087934
##	[331]	522.904369	2552.715451	24052.054569	1433.880588	7531.587612
##	[336]	8484.520189	4557.828693	3656.147052	6277.329558	2291.013936
##	[341]	6872.484780	493.894884	3015.094283	7897.151778	8634.307531
##	[346]	28155.212728	1012.265574	17764.608471	26924.839300	34062.041408
##	[351]	645.394815	22204.836896	19499.320096	23216.333512	15469.182391
##	[356]	7265.617215	956.550174	26779.794481	8443.273409	8808.448412
##	[361]	47428.130991	15693.018330	14719.114570	2762.416921	5450.403453
##	[366]	3253.830639	19378.963694	1635.363842	7254.351007	14050.668661
##	[371]	875.828921	2387.939955	7637.550651	2398.031123	19892.629484
##	[376]	10437.826402	31937.762695	6600.681591	26002.300967	9734.618677
##	[381]	27555.061302	634.432158	21191.957355	10749.119093	9129.594932
##	[386]	35400.521769	2209.572834	24223.245534	3042.545661	1025.267978
##	[391]	12402.126123	1385.405191	28566.035963	10638.992097	29271.173719
##	[396]	16844.100039	35085.423025	5904.078888	11749.955169	15913.839835
##	[401]	1754.612622	12730.313515	23837.108139	5985.088766	11408.800197
##	[406]	5938.296239	32725.791072	10435.290899	10282.548776	930.341632
##	[411]	16581.755918	7069.410528	310.577905	3207.181640	2775.471627
##	[416]	25479.885681	63.659437	5089.107378	11973.659869	2759.855872
##	[421]	3507.778195	6649.658577	2212.281342	10875.744980	20658.422695
##	[426]	6814.335778	4068.525080	12650.720789	9361.456893	10390.958457
##	[431]	2030.950648	611.370501	20506.047723	1396.989642	4461.314401
##	[436]	2206.387345	4930.306832	6143.639872	19060.817916	1119.846535
##	[441]	1276.532322	18202.651814	13730.568054	8243.136505	4416.221222
##	[446]	12534.611164	30233.155184	14137.894538	17267.424818	6125.916457
##	[451]	2008.825927	3564.113864	664.553774	2704.981183	6044.094997
##	[456]	15675.250550	30293.837786	1504.340746	1481.125097	6840.403279
##	[461]	40031.076585	11200.981863	5714.159680	511.863739	3337.978742
##	[466]	29386.779481	10353.723382	2524.603683	4857.879582	22515.860206
##	[471]	3727.165724	9736.827193	14264.048817	1479.265870	18122.865260
##	[476]	8892.012935	4320.213757	628.340525	416.298462	16849.932108
##	[481]	25293.633144	9759.032311	9310.422658	3462.314634	15085.505019
##	[486]	3409.496354	4212.448444	3371.852110	10100.478909	18146.576647
##	[491]	11668.748755	584.916406	3266.908426	4140.868840	28249.242568
##	[496]	3418.984693	2160.849900	12188.204356	45624.935639	28525.780829
##	[501]	15291.718028	213.273318	18549.265323	54356.939633	23888.293486
##	[506]	27686.955803	14400.149467	14429.034675	11858.964823	352.182070
##	[511]	937.436836	9070.657528	14752.097570	11328.281462	3284.316075
##	[516]	20734.949398	4269.710153	1974.145934	15515.961972	4597.355921
##	[521]	7890.627678	7159.567112	527.924661	1596.066075	1909.702017
##	[526]	6048.585055	3986.179693	11066.356588	27079.869239	37231.436941
##	[531]	292.890899	6457.729282	15542.461838	9102.932164	2860.762383
##	[536]	7437.031323	10406.754056	23929.568356	3058.351210	16442.823283
##	[541]	25207.956862	22024.019370	4688.554160	3123.872483	1150.617233
##	[546]	1445.827957	4847.352194	32.131666	41.953278	12857.940088
##	[551]	8597.401363	35682.037726	8510.870606	10372.542841	2329.747784
##	[556]	20884.213241	32329.234660	18595.174951	2518.883750	12817.296538
##	[561]	6285.986875	4121.294513	20795.035336	3895.243332	1054.885729
##	[566]	13185.506254	20784.527328	3180.835205	1685.340558	14849.322116
##	[571]	25002.300716	8871.562399	10253.808165	1842.937291	9604.905944

##	[576]	5169.952750	21460.272424	145.774810	24274.111633	2035.826812
##	[581]	3169.189328	525.543633	17797.112802	7905.057548	10172.551897
##	[586]	15441.912055	25968.949115	18657.797170	961.333750	582.410625
##	[591]	7581.989690	20122.879285	10471.654478	4379.169582	5573.332347
##	[596]	432.456674	11141.475176	4723.006422	4456.055831	109.590691
##	[601]	15961.549893	11367.261866	14557.707745	10854.381293	9859.720754
##	[606]	30888.043412	39396.256390	4267.743498	3038.644468	16502.078008
##	[611]	22106.019113	38221.339873	2886.041296	280.641102	6228.980858
##	[616]	4182.643894	230.397611	132.937428	29891.593774	4617.431325
##	[621]	3593.382377	47473.801498	25654.102920	3513.717164	21056.071797
##	[626]	8731.443970	7070.301373	2266.728068	9715.412574	14545.852855
##	[631]	839.357829	34404.194085	3082.362130	1104.226377	3735.500796
##	[636]	1151.453761	11286.924454	9336.885745	5577.999249	2122.859488
##	[641]	7316.372865	12752.315804	3418.688285	1439.528762	28676.331165
##	[646]	31214.759500	4555.051096	4598.942991	4556.082523	17195.591617
##	[651]	1343.252719	16128.430512	580.857212	969.558882	2767.502619
##	[656]	6315.161406	1583.309556	17259.156026	16283.188364	12915.619479
##	[661]	11843.703780	8881.092643	4881.556315	17619.725749	8501.749738
##	[666]	4856.600896	2.911364	5046.311392	2332.627936	7426.744026
##	[671]	11926.111778	15321.980815	14466.485535	12552.411745	779.646904
##	[676]	4478.838560	18748.082534	19019.056620	13070.373992	17278.044946
##	[681]	18322.813871	22635.580646	8948.071462	10979.083141	3096.925392
##	[686]	5193.259649	15579.340810	3024.716998	9571.714507	18435.146453
##	[691]	16480.255018	5087.693387	120.246009	178.514045	10802.717650
##	[696]	18459.186131	6597.060018	8513.922694	3587.560764	12126.023322
##	[701]	17694.996431	2938.394793	950.896817	13247.096773	20373.894881
##	[706]	5178.766812	15272.573192	301.345414	10684.385570	7065.900588
##	[711]	14970.625787	12601.051367	18815.045207	12519.580012	1416.591695
##	[716]	3073.618144	8617.932564	12467.393741	7926.797834	13327.189915
##	[721]	8882.203827	3101.915562	1933.480141	38004.972763	5482.761977
##	[726]	10693.552414	4521.717473	17461.485485	1168.490829	6969.097734
##	[731]	208.900800	4954.955769	8616.033620	8275.692480	17849.116282
##	[736]	1867.267262	30137.003165	1709.512964	2153.543932	18773.086670
##	[741]	16314.069761	13750.959121	275.013705	7029.379923	13753.425142
##	[746]	3585.834138	1.586600	4621.954548	24180.794951	10589.006559
##	[751]	31699.762048	73437.226397	1014.720572	8268.486899	6392.931115
##	[756]	29936.270610	8405.116192	4531.588439	17023.508672	1285.091553
##	[761]	4444.785987	5360.671090	10592.690847	892.674683	5460.578925
##	[766]	1118.332614	561.104135	1407.046838	20761.447255	14364.669702
##	[771]	54119.344072	1411.943890	12517.124128	1728.615848	165.268369
##	[776]	15141.414672	17869.620637	6672.821515	18712.423570	7780.916923
##	[781]	26283.493800	2449.484568	8516.893361	8618.388938	24779.255670
##	[786]	1473.913482	6432.769463	1116.148640	556.425791	2927.925506
##	[791]	2198.784859	2587.054083	16023.730000	1012.590991	11961.781800
##	[796]	1968.475175	1565.745108	6362.282280	2075.313162	5321.446647
##	[801]	773.033255	14764.661576	2506.478131	13212.812524	1109.514049
##	[806]	1392.721043	8420.662271	3603.200593	2801.796365	3875.457040
##	[811]	183.752519	24166.483876	2841.846378	5295.602282	16937.724502
##	[816]	1352.773561	21319.562588	5721.908413	7094.807690	10675.368802
##	[821]	13911.707117	1043.289915	10601.815949	334.580850	954.449267
##	[826]	16964.339184	4495.326125	6688.084758	2412.930947	2701.402732
##	[831]	199.976920	9554.287608	7430.985615	13588.927263	3138.215206
##	[836]	9912.594294	21256.511314	1322.030306	13771.886616	3853.706654
##	[841]	10206.481802	2854.539248	10560.178697	9973.180646	14116.044204

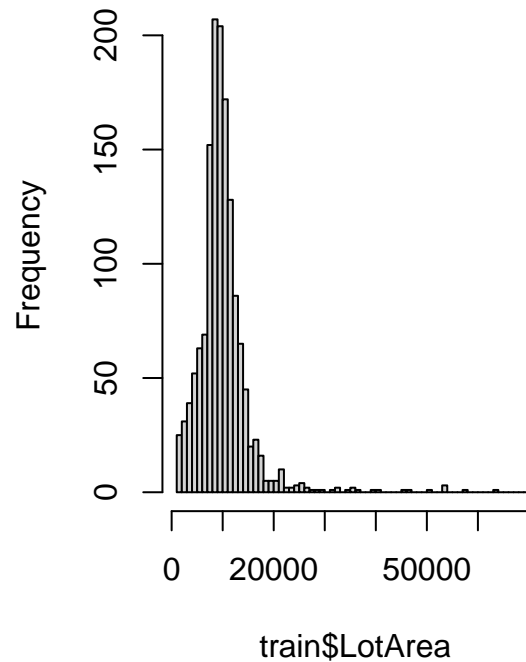
```
## [846] 6468.035697 19150.509982 33101.187081 1210.440558 2779.834853
## [851] 276.752890 18029.593601 9022.090357 2675.694459 30684.362082
## [856] 9705.833556 162.440016 7889.661516 12805.442475 7463.632228
## [861] 1168.946221 17359.059433 11118.713380 16437.663840 13188.256916
## [866] 3292.694491 5717.583784 2537.258824 343.419191 12862.025169
## [871] 21967.475676 7385.886516 11036.181430 22030.804237 12115.082523
## [876] 20299.975922 18803.358577 21887.991023 33886.359503 13398.315192
## [881] 2571.171738 36469.019575 18533.860291 1635.371059 16455.316158
## [886] 6240.058332 16831.139325 41497.564895 14948.806706 6720.791142
## [891] 10272.720368 15554.134545 3145.449345 7013.390277 440.769955
## [896] 5586.995825 1185.429967 16762.909013 106.540696 8763.591278
## [901] 1553.421129 19818.313582 12017.297919 16516.567667 7235.586085
## [906] 17637.960504 16247.287439 14162.653648 6614.473516 80.966591
## [911] 3747.529014 871.149600 5981.543438 1412.705316 27668.575491
## [916] 18726.208141 11440.247407 5661.233474 2541.202250 1886.261211
## [921] 147.503558 2039.480723 11261.454739 14183.088459 1143.534333
## [926] 5676.761640 4689.618007 325.489020 12224.663393 3118.352817
## [931] 4188.295635 32245.914947 7032.485374 26245.669997 9658.412941
## [936] 944.598658 3827.160688 8607.047364 2667.189787 12183.824982
## [941] 9108.130478 27005.701493 26832.882533 325.080742 3052.047260
## [946] 8684.330930 12800.381836 47342.924714 6257.082340 2331.766080
## [951] 8311.747013 4202.287169 3700.057921 10287.866055 1494.454753
## [956] 5775.005742 5470.158014 500.304736 16079.600450 2640.780587
## [961] 7302.092585 10655.770648 1553.983571 2932.010000 1452.289621
## [966] 18072.862949 4380.741420 2040.133562 437.396406 20798.786536
## [971] 4233.884768 23780.180272 8332.681846 45807.021602 2610.010970
## [976] 1475.611905 2661.326989 1272.384199 3443.817813 3998.445513
## [981] 9128.095889 2709.427494 26546.407486 2083.494454 49552.561789
## [986] 23216.518979 25374.126515 8078.335234 5605.701484 5473.094571
## [991] 5700.056701 26616.571202 13093.457779 5956.223062 60.493708
## [996] 10885.181493 8598.745918 15722.304956 30105.551627 7389.696597
```

```
# Plot histogram of original and exponential
par(mfrow = c(1, 2))
hist(samples, breaks = 50, main = "Exponential Lot Area")
hist(train$LotArea, breaks = 50, main = "Original Lot Area")
```

Exponential Lot Area



Original Lot Area



```
# Percentiles based on exponential
qexp(0.05, rate = lambda)
```

```
## [1] 517.8831
```

```
qexp(0.95, rate = lambda)
```

```
## [1] 30246.43
```

```
# Empirical 95% confidence interval assuming normality
me <- qnorm(0.975)*nrow(train)/sqrt(sd(train$LotArea))
upper <- mean(train$LotArea) + me
lower <- mean(train$LotArea) - me
```

```
# Empirical percentiles
quantile(train$LotArea, 0.05)
```

```
##      5%
## 3294.5
```

```
quantile(train$LotArea, 0.95)
```

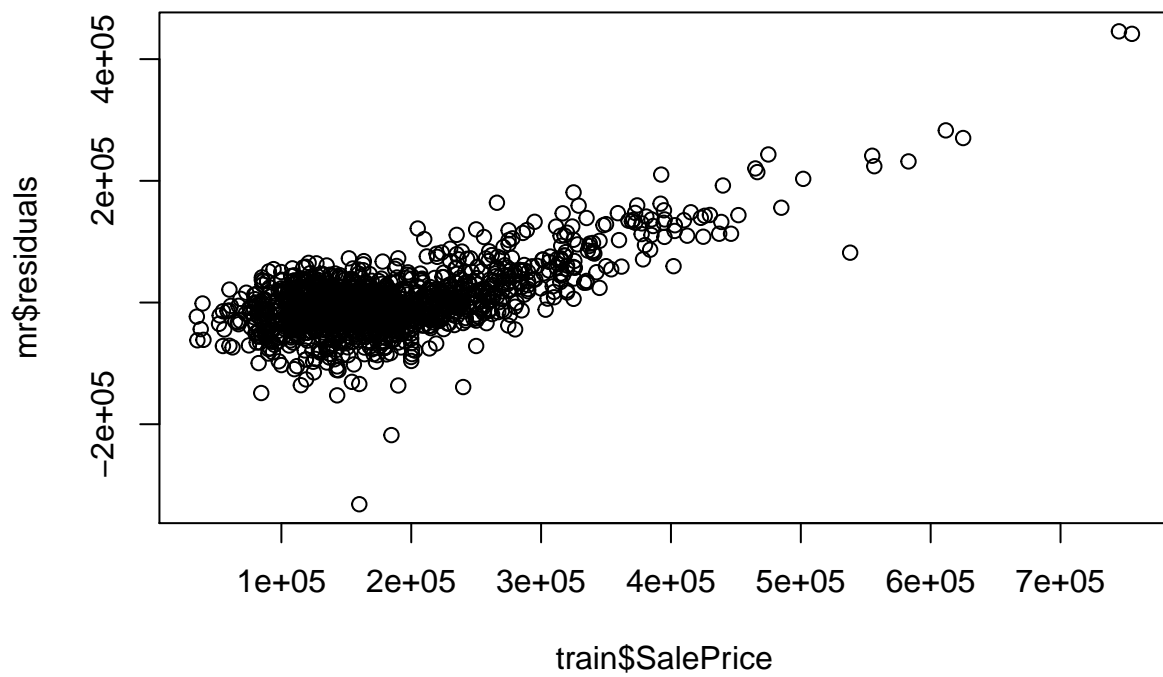
```
##     95%
## 17108
```

Modeling

```
# Create multiple regression
mr <- lm(SalePrice ~ LotArea + YearBuilt + TotRmsAbvGrd, data = train)
summary(mr)

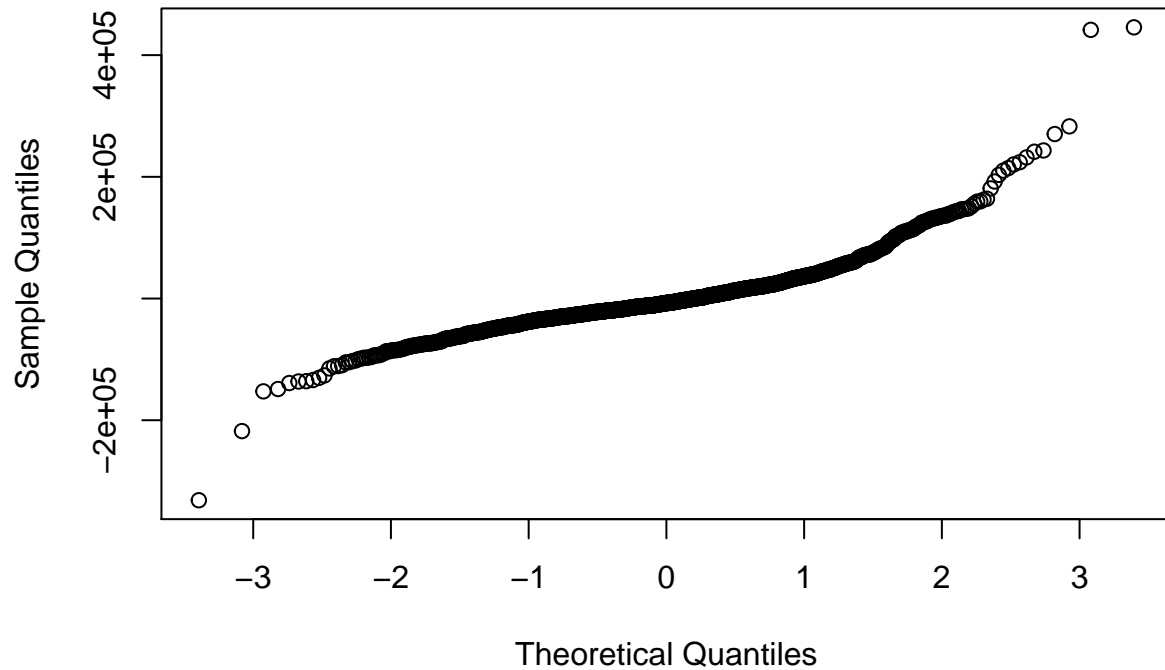
##
## Call:
## lm(formula = SalePrice ~ LotArea + YearBuilt + TotRmsAbvGrd,
##     data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -331486  -27636   -7406   19341  445648
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.444e+06  9.110e+04  -26.83  <2e-16 ***
## LotArea      2.809e+00  2.605e-01   10.79  <2e-16 ***
## YearBuilt    1.248e+03  4.640e+01   26.91  <2e-16 ***
## TotRmsAbvGrd 2.078e+04  9.080e+02   22.88  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 53270 on 1452 degrees of freedom
## Multiple R-squared:  0.5494, Adjusted R-squared:  0.5484
## F-statistic: 590 on 3 and 1452 DF, p-value: < 2.2e-16

plot(mr$residuals ~ train$SalePrice)
```

```
qqnorm(mr$residuals)
```

Normal Q-Q Plot



```
submission <- (2.809*test$LotArea)+(1248*test$YearBuilt)+(20780*test$TotRmsAbvGrd)-2444000  
  
#submission <- cbind(test,submission) %>% select(Id, submission)  
#names(submission) <- c("Id", "SalePrice")  
#write.csv(submission, file = "DMosteSubmission.csv")
```

My username is davidmoste and I hade a score of 0.285.