

# DS311 - R Lab Assignment

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## R Assignment 1

- In this assignment, we are going to apply some of the built-in data sets in R for descriptive statistics analysis.
- To earn full grade in this assignment, students need to complete the coding tasks for each question to get the result.
- After finishing all the questions, knit the document into HTML format for submission.

### Question 1

Using the **mtcars** data set in R, please answer the following questions.

```
# Loading the data  
data(mtcars)
```

```
# Head of the data set  
head(mtcars)
```

```
##           mpg  cyl  disp  hp  drat    wt    qsec vs  am  gear  carb  
## Mazda RX4      21.0   6  160 110 3.90 2.620 16.46  0  1    4    4  
## Mazda RX4 Wag  21.0   6  160 110 3.90 2.875 17.02  0  1    4    4  
## Datsun 710      22.8   4  108  93 3.85 2.320 18.61  1  1    4    1  
## Hornet 4 Drive  21.4   6  258 110 3.08 3.215 19.44  1  0    3    1  
## Hornet Sportabout 18.7   8  360 175 3.15 3.440 17.02  0  0    3    2  
## Valiant        18.1   6  225 105 2.76 3.460 20.22  1  0    3    1
```

a. Report the number of variables and observations in the data set.

```
# Enter your code here!  
?mtcars
```

```
## starting httpd help server ... done
```

```
dim(mtcars)
```

```
## [1] 32 11
```

```
names(mtcars)
```

```
## [1] "mpg" "cyl" "disp" "hp" "drat" "wt" "qsec" "vs" "am" "gear"  
## [11] "carb"
```

```
# Answer:
```

```
print("There are total of 11 numeric variables and 32 observations in this data set.")
```

```
## [1] "There are total of 11 numeric variables and 32 observations in this data set."
```

- b. Print the summary statistics of the data set and report how many discrete and continuous variables are in the data set.

```
# Enter your code here!
```

```
summary(mtcars)
```

```
##      mpg          cyl          disp          hp  
##  Min.   :10.40   Min.   :4.000   Min.   : 71.1   Min.   : 52.0  
## 1st Qu.:15.43   1st Qu.:4.000   1st Qu.:120.8   1st Qu.: 96.5  
## Median :19.20   Median :6.000   Median :196.3   Median :123.0  
## Mean   :20.09   Mean   :6.188   Mean   :230.7   Mean   :146.7  
## 3rd Qu.:22.80   3rd Qu.:8.000   3rd Qu.:326.0   3rd Qu.:180.0  
## Max.   :33.90   Max.   :8.000   Max.   :472.0   Max.   :335.0  
##      drat          wt          qsec          vs  
##  Min.   :2.760   Min.   :1.513   Min.   :14.50   Min.   :0.0000  
## 1st Qu.:3.080   1st Qu.:2.581   1st Qu.:16.89   1st Qu.:0.0000  
## Median :3.695   Median :3.325   Median :17.71   Median :0.0000  
## Mean   :3.597   Mean   :3.217   Mean   :17.85   Mean   :0.4375  
## 3rd Qu.:3.920   3rd Qu.:3.610   3rd Qu.:18.90   3rd Qu.:1.0000  
## Max.   :4.930   Max.   :5.424   Max.   :22.90   Max.   :1.0000  
##      am          gear          carb  
##  Min.   :0.0000   Min.   :3.000   Min.   :1.000  
## 1st Qu.:0.0000   1st Qu.:3.000   1st Qu.:2.000  
## Median :0.0000   Median :4.000   Median :2.000  
## Mean   :0.4062   Mean   :3.688   Mean   :2.812  
## 3rd Qu.:1.0000   3rd Qu.:4.000   3rd Qu.:4.000  
## Max.   :1.0000   Max.   :5.000   Max.   :8.000
```

```
# Answer:
```

```
print("There are 5 discrete variables and 6 continuous variables in this data set.")
```

```
## [1] "There are 5 discrete variables and 6 continuous variables in this data set."
```

- c. Calculate the mean, variance, and standard deviation for the variable **mpg** and assign them into variable names **m**, **v**, and **s**. Report the results in the print statement.

```
# Enter your code here!
```

```
m <- mean(mtcars$mpg)
```

```
v<-var(mtcars$mpg)
```

```
s<-sd(mtcars$mpg)
```

```
print(paste("The average of Mile Per Gallon from this data set is ", m , " with variance ", v , " and s
```

```
## [1] "The average of Mile Per Gallon from this data set is 20.090625 with variance 36.324102822580
```

- d. Create two tables to summarize 1) average mpg for each cylinder class and 2) the standard deviation of mpg for each gear class.

```
# Enter your code here!
m_mpg_cyl<-aggregate(mpg ~ cyl, mtcars, mean)
sd_mpg_gr<-aggregate(mpg ~ cyl, mtcars, sd)

print(m_mpg_cyl)
```

```
##   cyl      mpg
## 1    4 26.66364
## 2    6 19.74286
## 3    8 15.10000
```

```
print(sd_mpg_gr)
```

```
##   cyl      mpg
## 1    4 4.509828
## 2    6 1.453567
## 3    8 2.560048
```

- e. Create a crosstab that shows the number of observations belong to each cylinder and gear class combinations. The table should show how many observations given the car has 4 cylinders with 3 gears, 4 cylinders with 4 gears, etc. Report which combination is recorded in this data set and how many observations for this type of car.

```
# Enter your code here!
cyl_gr_htable<-table(mtcars$cyl, mtcars$gear)
print(cyl_gr_htable)
```

```
##
##      3  4  5
## 4   1  8  2
## 6   2  4  1
## 8  12  0  2
```

```
print("The most common car type in this data set is car with 8 cylinders and 3 gears. There are total o
```

```
## [1] "The most common car type in this data set is car with 8 cylinders and 3 gears. There are total o
```

## Question 2

Use different visualization tools to summarize the data sets in this question.

- Using the **PlantGrowth** data set, visualize and compare the weight of the plant in the three separated group. Give labels to the title, x-axis, and y-axis on the graph. Write a paragraph to summarize your findings.

```
# Load the data set  
data("PlantGrowth")
```

```
# Head of the data set  
head(PlantGrowth)
```

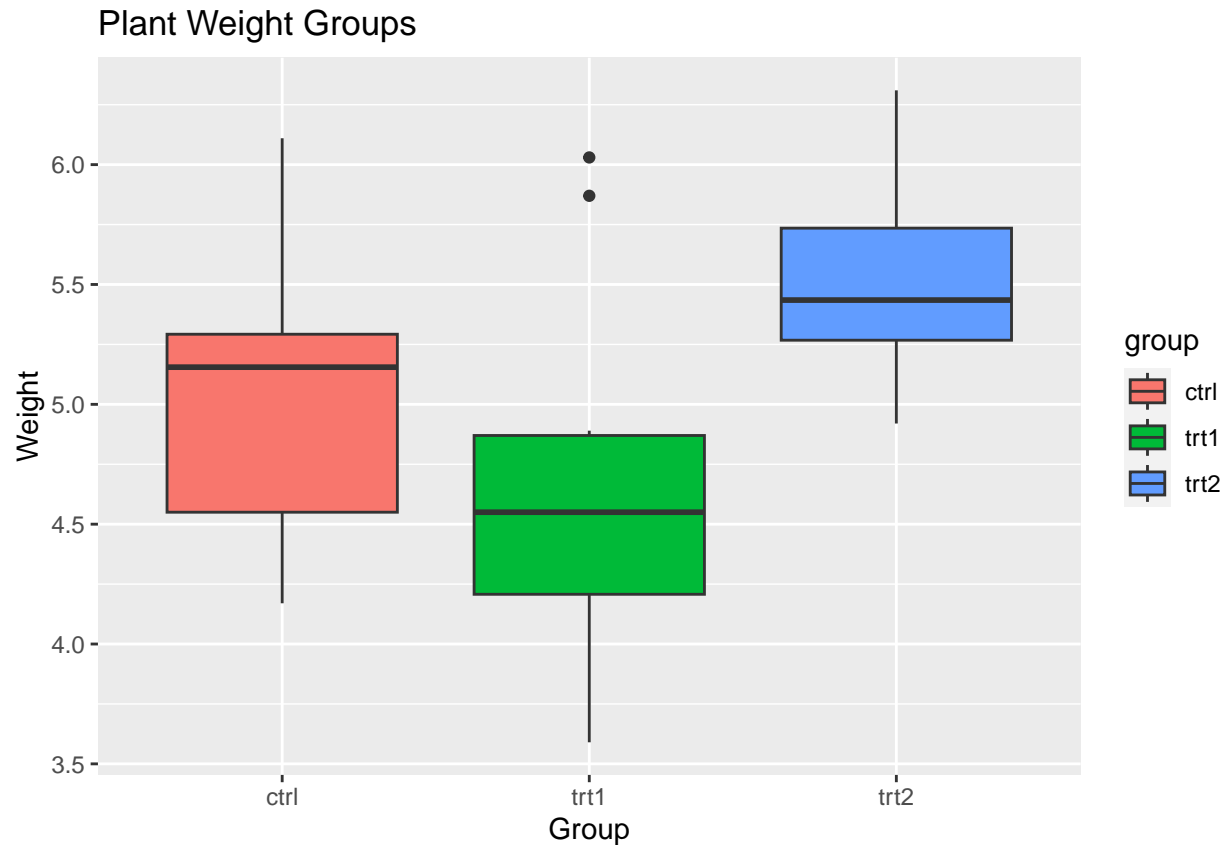
```
##   weight group  
## 1   4.17  ctrl  
## 2   5.58  ctrl  
## 3   5.18  ctrl  
## 4   6.11  ctrl  
## 5   4.50  ctrl  
## 6   4.61  ctrl
```

```
# Enter your code here!
```

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 4.3.2
```

```
ggplot(PlantGrowth, aes(x=group, y=weight, fill=group))+geom_boxplot()+  
  labs(title="Plant Weight Groups", x="Group", y="Weight")
```

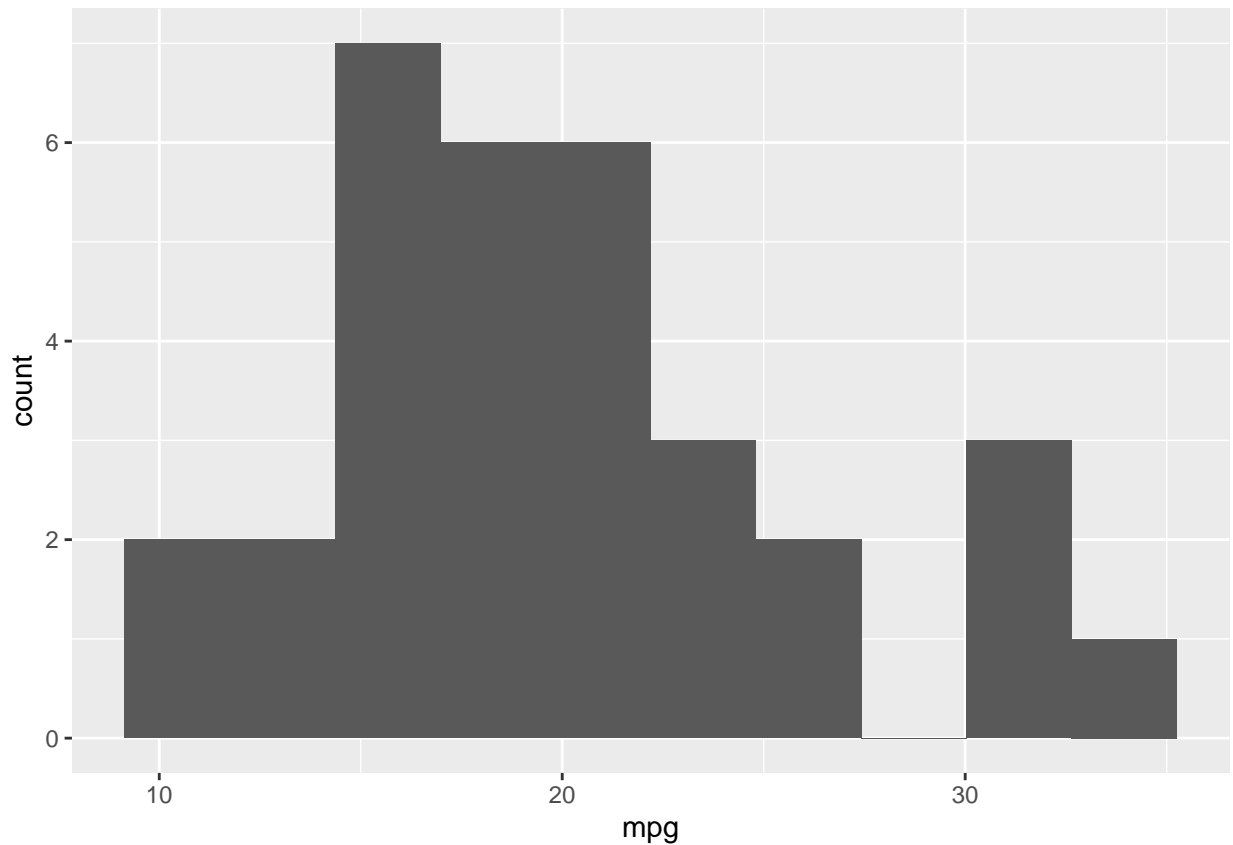


Result:

=> Report a paragraph to summarize your findings from the plot! The Plant weight groups reflect that the treatment groups provided vastly different results on both ends of the control group. Treatment 1 had the most similarities with the control group while treatment 2 was on the upper end of the control range barely intersecting at all. Treatment 2 produced bigger, healthier plants given this data. Treatment 2 had the highest median followed by the control and then treatment 1. The upper quartile for treatment 1 was very small with most plants in this group reflecting quartile 3 and 2. the median for the control group was roughly ~5.2, treatment 1 ~4.55, and treatment 3 at ~5.4.

- b. Using the **mtcars** data set, plot the histogram for the column **mpg** with 10 breaks. Give labels to the title, x-axis, and y-axis on the graph. Report the most observed mpg class from the data set.

```
ggplot(data=mtcars)+geom_histogram(aes(x=mpg),bins=10)
```



```
print("Most of the cars in this data set are in the class of 15 mile per gallon.")
```

```
## [1] "Most of the cars in this data set are in the class of 15 mile per gallon."
```

- c. Using the **USArrests** data set, create a pairs plot to display the correlations between the variables in the data set. Plot the scatter plot with **Murder** and **Assault**. Give labels to the title, x-axis, and y-axis on the graph. Write a paragraph to summarize your results from both plots.

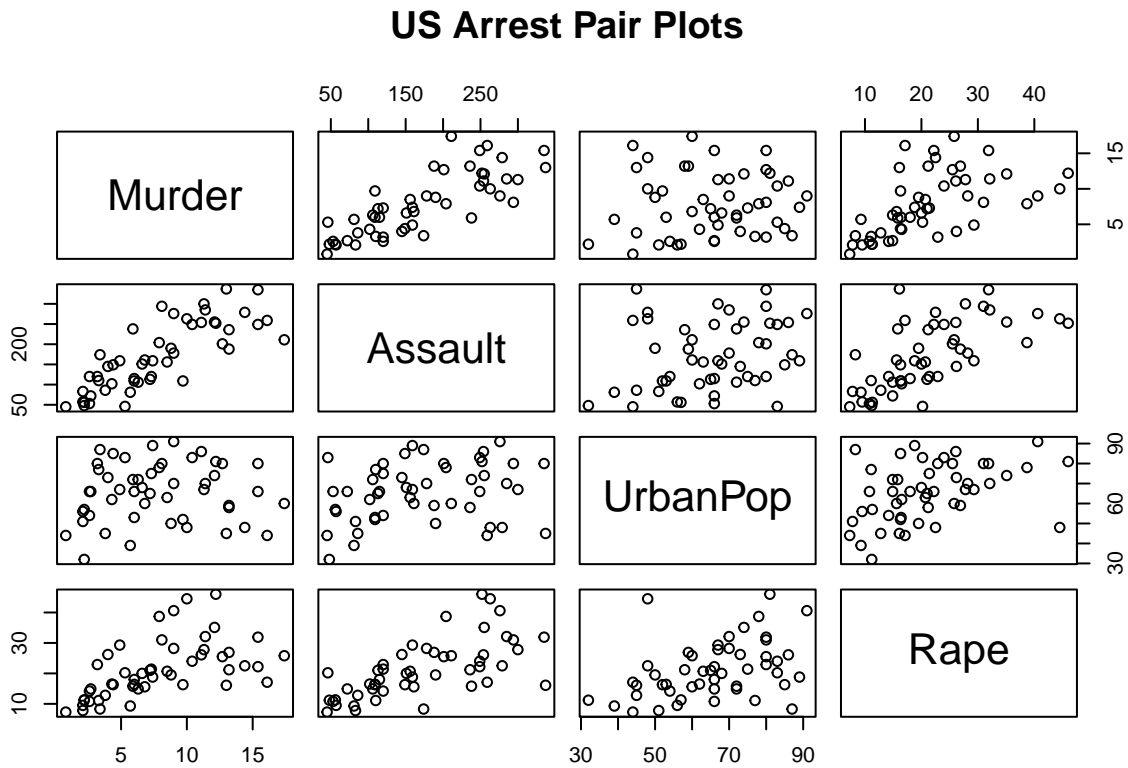
```
# Load the data set
data("USArrests")

# Head of the data set
head(USArrests)
```

```
##           Murder  Assault  UrbanPop  Rape
## Alabama      13.2     236       58 21.2
## Alaska       10.0     263       48 44.5
## Arizona       8.1     294       80 31.0
## Arkansas      8.8     190       50 19.5
## California    9.0     276       91 40.6
## Colorado     7.9     204       78 38.7
```

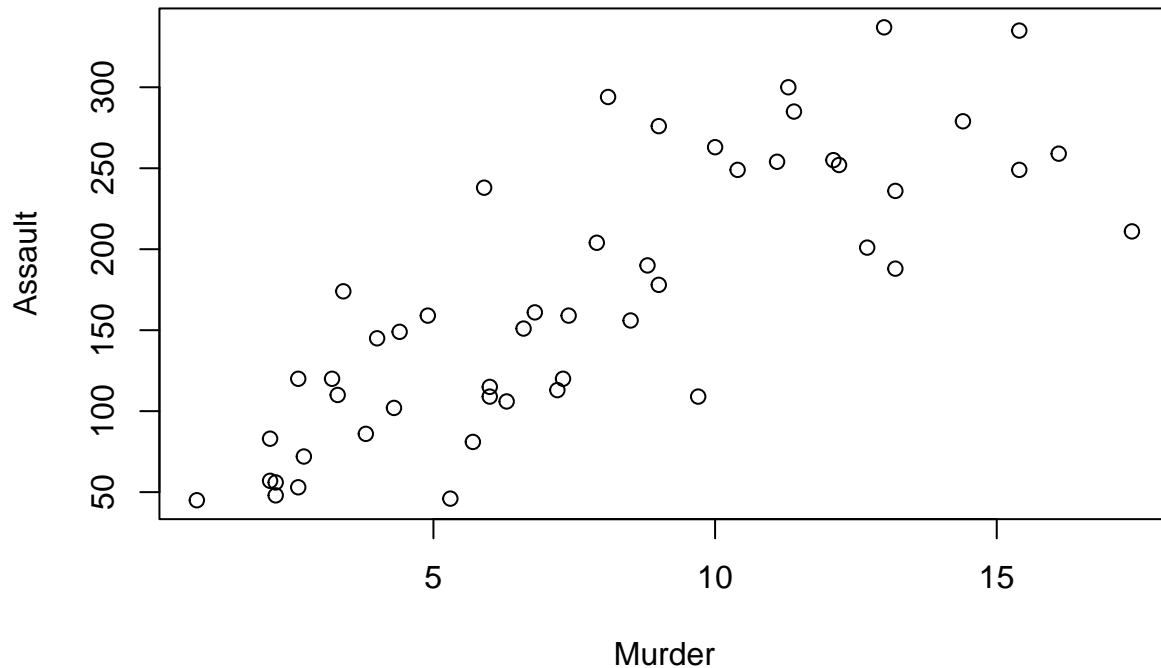
*# Enter your code here!*

```
pairs(USArrests, main="US Arrest Pair Plots")
```



```
plot(USArrests$Murder, USArrests$Assault, xlab= "Murder", ylab="Assault", main="Murder & Assault Rates")
```

## Murder & Assault Rates in US



Result:

=> Report a paragraph to summarize your findings from the plot!

Assaults are far more common than murder. As assaults increase, murders also increase showing a positive correlation. The trend also seems pretty linear in nature.

---

### Question 3

Download the housing data set from [www.jaredlander.com](http://www.jaredlander.com) and find out what explains the housing prices in New York City.

Note: Check your working directory to make sure that you can download the data into the data folder.

- Create your own descriptive statistics and aggregation tables to summarize the data set and find any meaningful results between different variables in the data set.

```
# Head of the cleaned data set  
head(housingData)
```

```
##   Neighborhood Market.Value.per.SqFt      Boro Year.Built  
## 1    FINANCIAL          200.00 Manhattan    1920  
## 2    FINANCIAL          242.76 Manhattan    1985  
## 4    FINANCIAL          271.23 Manhattan    1930
```



```
## 5      TRIBECA      247.48 Manhattan      1985
## 6      TRIBECA      191.37 Manhattan      1986
## 7      TRIBECA      211.53 Manhattan      1985
```

*# Enter your code here!*

```
summary(housingData)
```

```
## Neighborhood      Market.Value.per.SqFt      Boro      Year.Built
## Length:2530      Min.      : 10.66      Length:2530      Min.      :1825
## Class :character  1st Qu.: 75.10      Class :character  1st Qu.:1926
## Mode  :character  Median :114.89      Mode  :character  Median :1986
##                      Mean      :133.17                      Mean      :1967
##                      3rd Qu.:189.91                      3rd Qu.:2005
##                      Max.      :399.38                      Max.      :2010
```

```
MVSFmean<-aggregate(Market.Value.per.SqFt~Boro, data=housingData, FUN=mean)
MVSFvar<-aggregate(Market.Value.per.SqFt~Boro, data=housingData, FUN=var)
MVSFsd<-aggregate(Market.Value.per.SqFt~Boro, data=housingData, FUN=sd)
```

```
print(MVSFmean)
```

```
##      Boro Market.Value.per.SqFt
## 1      Bronx      47.93232
## 2      Brooklyn      80.13439
## 3      Manhattan      180.59265
## 4      Queens      77.38137
## 5 Staten Island      41.26958
```

```
print(MVSFvar)
```

```
##      Boro Market.Value.per.SqFt
## 1      Bronx      227.3035
## 2      Brooklyn      524.9035
## 3      Manhattan      3146.9929
## 4      Queens      947.1520
## 5 Staten Island      46.1433
```

```
print(MVSFsd)
```

```
##      Boro Market.Value.per.SqFt
## 1      Bronx      15.076589
## 2      Brooklyn      22.910773
## 3      Manhattan      56.098065
## 4      Queens      30.775835
## 5 Staten Island      6.792886
```

```
MVYBmean<-aggregate(Market.Value.per.SqFt~Year.Built, data=housingData, FUN=mean)
MVYBvar<-aggregate(Market.Value.per.SqFt~Year.Built, data=housingData, FUN=var)
MVYBsd<-aggregate(Market.Value.per.SqFt~Year.Built, data=housingData, FUN=sd)
```

```
print(MVYBmean)
```

##	Year.Built	Market.Value.per.SqFt
## 1	1825	76.36000
## 2	1836	273.77000
## 3	1853	152.79000
## 4	1860	159.64500
## 5	1874	111.17000
## 6	1875	166.05000
## 7	1879	194.52000
## 8	1881	109.70500
## 9	1883	172.10000
## 10	1890	113.28750
## 11	1891	72.83000
## 12	1892	95.21000
## 13	1893	168.85000
## 14	1894	110.62000
## 15	1895	151.77500
## 16	1896	117.26500
## 17	1897	40.83000
## 18	1898	83.25000
## 19	1899	108.16000
## 20	1900	137.55908
## 21	1901	172.36778
## 22	1902	167.62167
## 23	1903	147.97000
## 24	1904	123.09333
## 25	1905	187.76583
## 26	1906	169.03364
## 27	1907	173.80000
## 28	1908	150.35000
## 29	1909	135.23667
## 30	1910	147.36257
## 31	1911	179.76067
## 32	1912	159.51636
## 33	1913	175.93500
## 34	1914	160.29286
## 35	1915	147.08673
## 36	1916	128.20714
## 37	1917	73.87000
## 38	1918	181.84000
## 39	1919	63.11000
## 40	1920	145.30862
## 41	1921	122.39125
## 42	1922	118.33250
## 43	1923	115.47625
## 44	1924	165.94091
## 45	1925	147.51316
## 46	1926	148.36423
## 47	1927	131.63357
## 48	1928	153.68375
## 49	1929	106.32121
## 50	1930	142.28936
## 51	1931	129.51731
## 52	1932	91.74333
## 53	1933	40.97000

## 54	1934	203.80000
## 55	1935	176.23000
## 56	1936	46.04333
## 57	1937	51.77250
## 58	1938	99.23857
## 59	1939	93.65083
## 60	1940	154.89857
## 61	1941	111.83733
## 62	1942	128.38600
## 63	1947	113.13500
## 64	1948	186.25000
## 65	1949	44.98000
## 66	1950	141.96182
## 67	1951	132.98833
## 68	1952	97.95143
## 69	1954	81.56500
## 70	1955	130.17538
## 71	1956	178.42786
## 72	1957	127.24091
## 73	1958	159.77000
## 74	1959	108.62692
## 75	1960	104.91200
## 76	1961	106.63000
## 77	1962	129.26294
## 78	1963	152.82937
## 79	1964	103.15000
## 80	1965	121.01313
## 81	1966	79.94375
## 82	1967	91.94000
## 83	1968	126.76000
## 84	1969	157.28000
## 85	1970	214.59000
## 86	1971	57.60000
## 87	1972	185.72000
## 88	1973	196.75500
## 89	1974	124.42500
## 90	1975	201.26667
## 91	1977	161.32250
## 92	1978	254.69000
## 93	1979	155.71333
## 94	1980	161.74500
## 95	1981	175.96800
## 96	1982	151.30364
## 97	1983	114.79917
## 98	1984	179.48333
## 99	1985	182.66868
## 100	1986	157.62328
## 101	1987	142.14055
## 102	1988	126.43686
## 103	1989	109.25390
## 104	1990	99.31500
## 105	1991	145.76105
## 106	1992	83.92333
## 107	1993	55.45000

## 108	1994	73.13500
## 109	1995	75.77375
## 110	1996	152.36750
## 111	1997	137.41364
## 112	1998	138.25125
## 113	1999	145.93217
## 114	2000	165.47296
## 115	2001	124.74295
## 116	2002	117.92442
## 117	2003	121.56193
## 118	2004	113.79702
## 119	2005	122.70817
## 120	2006	119.73598
## 121	2007	134.12665
## 122	2008	144.34935
## 123	2009	96.52619
## 124	2010	90.36667

```
print(MVYBvar)
```

##	Year.Built	Market.Value.per.SqFt
## 1	1825	NA
## 2	1836	NA
## 3	1853	NA
## 4	1860	3258.666450
## 5	1874	NA
## 6	1875	4523.907200
## 7	1879	0.000000
## 8	1881	530.402450
## 9	1883	NA
## 10	1890	1368.484825
## 11	1891	NA
## 12	1892	NA
## 13	1893	210.125000
## 14	1894	474.936200
## 15	1895	2060.178050
## 16	1896	2483.010450
## 17	1897	NA
## 18	1898	NA
## 19	1899	254.218900
## 20	1900	2716.073519
## 21	1901	2381.137494
## 22	1902	2614.890137
## 23	1903	7867.772743
## 24	1904	5116.211187
## 25	1905	2499.019790
## 26	1906	6565.315225
## 27	1907	6351.736060
## 28	1908	3073.581040
## 29	1909	5231.293747
## 30	1910	3950.707225
## 31	1911	3272.755964
## 32	1912	2638.376824
## 33	1913	3197.830642

## 34	1914	6420.206124
## 35	1915	1610.568502
## 36	1916	2994.437724
## 37	1917	15.456800
## 38	1918	NA
## 39	1919	NA
## 40	1920	3756.092403
## 41	1921	2509.008813
## 42	1922	4730.988130
## 43	1923	3883.373884
## 44	1924	3318.562269
## 45	1925	3276.497145
## 46	1926	4403.390081
## 47	1927	6225.151563
## 48	1928	4614.811816
## 49	1929	4420.556698
## 50	1930	4208.858741
## 51	1931	4071.365388
## 52	1932	6129.611907
## 53	1933	374.522950
## 54	1934	NA
## 55	1935	NA
## 56	1936	2.104633
## 57	1937	645.077936
## 58	1938	5686.392981
## 59	1939	5181.063045
## 60	1940	3650.905681
## 61	1941	5302.484250
## 62	1942	5352.763680
## 63	1947	8492.954450
## 64	1948	1497.139200
## 65	1949	NA
## 66	1950	11349.892396
## 67	1951	10220.160137
## 68	1952	3400.728848
## 69	1954	2495.711250
## 70	1955	5940.094144
## 71	1956	3423.359649
## 72	1957	4052.401929
## 73	1958	3266.229900
## 74	1959	4968.902790
## 75	1960	5215.848089
## 76	1961	6241.680876
## 77	1962	3364.996822
## 78	1963	2937.430966
## 79	1964	4104.748978
## 80	1965	5268.727596
## 81	1966	1945.375284
## 82	1967	5328.842571
## 83	1968	16006.183200
## 84	1969	4120.063000
## 85	1970	2620.880000
## 86	1971	NA
## 87	1972	526.825800

## 88	1973	3727.558550
## 89	1974	6998.628050
## 90	1975	108.480433
## 91	1977	5467.827692
## 92	1978	194.439200
## 93	1979	7735.403433
## 94	1980	7562.697310
## 95	1981	4973.035084
## 96	1982	7212.187853
## 97	1983	5247.755251
## 98	1984	6428.969261
## 99	1985	5291.172644
## 100	1986	6028.058382
## 101	1987	7639.721625
## 102	1988	5561.143562
## 103	1989	5939.763334
## 104	1990	3970.088329
## 105	1991	6200.911421
## 106	1992	5694.161267
## 107	1993	336.823600
## 108	1994	194.587710
## 109	1995	4108.392055
## 110	1996	10250.771692
## 111	1997	8088.192905
## 112	1998	7845.024158
## 113	1999	5529.888081
## 114	2000	7727.011606
## 115	2001	5202.246635
## 116	2002	4201.514013
## 117	2003	3690.549708
## 118	2004	3349.703398
## 119	2005	4368.012474
## 120	2006	3345.778459
## 121	2007	4816.248152
## 122	2008	4926.957866
## 123	2009	4305.694575
## 124	2010	1128.816133

```
print(MVYBsd)
```

##	Year.Built	Market.Value.per.SqFt
## 1	1825	NA
## 2	1836	NA
## 3	1853	NA
## 4	1860	57.084730
## 5	1874	NA
## 6	1875	67.259997
## 7	1879	0.000000
## 8	1881	23.030468
## 9	1883	NA
## 10	1890	36.993038
## 11	1891	NA
## 12	1892	NA
## 13	1893	14.495689

## 14	1894	21.793031
## 15	1895	45.389184
## 16	1896	49.829815
## 17	1897	NA
## 18	1898	NA
## 19	1899	15.944243
## 20	1900	52.115962
## 21	1901	48.796900
## 22	1902	51.135996
## 23	1903	88.700466
## 24	1904	71.527695
## 25	1905	49.990197
## 26	1906	81.026633
## 27	1907	79.697780
## 28	1908	55.439887
## 29	1909	72.327683
## 30	1910	62.854652
## 31	1911	57.208006
## 32	1912	51.365132
## 33	1913	56.549365
## 34	1914	80.126189
## 35	1915	40.131889
## 36	1916	54.721456
## 37	1917	3.931514
## 38	1918	NA
## 39	1919	NA
## 40	1920	61.286968
## 41	1921	50.090007
## 42	1922	68.782179
## 43	1923	62.316722
## 44	1924	57.606964
## 45	1925	57.240695
## 46	1926	66.358045
## 47	1927	78.899630
## 48	1928	67.932406
## 49	1929	66.487267
## 50	1930	64.875718
## 51	1931	63.807252
## 52	1932	78.291838
## 53	1933	19.352595
## 54	1934	NA
## 55	1935	NA
## 56	1936	1.450735
## 57	1937	25.398385
## 58	1938	75.408176
## 59	1939	71.979602
## 60	1940	60.422725
## 61	1941	72.818159
## 62	1942	73.162584
## 63	1947	92.157227
## 64	1948	38.692883
## 65	1949	NA
## 66	1950	106.535874
## 67	1951	101.094808

## 68	1952	58.315768
## 69	1954	49.957094
## 70	1955	77.072006
## 71	1956	58.509483
## 72	1957	63.658479
## 73	1958	57.150940
## 74	1959	70.490445
## 75	1960	72.220829
## 76	1961	79.004309
## 77	1962	58.008593
## 78	1963	54.198072
## 79	1964	64.068315
## 80	1965	72.586001
## 81	1966	44.106409
## 82	1967	72.998922
## 83	1968	126.515545
## 84	1969	64.187717
## 85	1970	51.194531
## 86	1971	NA
## 87	1972	22.952686
## 88	1973	61.053735
## 89	1974	83.657803
## 90	1975	10.415394
## 91	1977	73.944761
## 92	1978	13.944146
## 93	1979	87.951142
## 94	1980	86.963770
## 95	1981	70.519750
## 96	1982	84.924601
## 97	1983	72.441392
## 98	1984	80.180853
## 99	1985	72.740447
## 100	1986	77.640572
## 101	1987	87.405501
## 102	1988	74.573075
## 103	1989	77.069860
## 104	1990	63.008637
## 105	1991	78.745866
## 106	1992	75.459666
## 107	1993	18.352755
## 108	1994	13.949470
## 109	1995	64.096740
## 110	1996	101.246095
## 111	1997	89.934381
## 112	1998	88.572141
## 113	1999	74.363217
## 114	2000	87.903422
## 115	2001	72.126601
## 116	2002	64.819087
## 117	2003	60.749895
## 118	2004	57.876622
## 119	2005	66.090941
## 120	2006	57.842704
## 121	2007	69.399194



```
## 122      2008      70.192292
## 123      2009      65.617792
## 124      2010      33.597859
```

```
MVNmean<-aggregate(Market.Value.per.SqFt~Neighborhood, data=housingData, FUN=mean)
MVNvar<-aggregate(Market.Value.per.SqFt~Neighborhood, data=housingData, FUN=var)
MVNsd<-aggregate(Market.Value.per.SqFt~Neighborhood, data=housingData, FUN=sd)

print(MVNmean)
```

```
##      Neighborhood Market.Value.per.SqFt
## 1      ALPHABET CITY      148.35500
## 2  ARROCHAR-SHORE ACRES      57.75000
## 3      ASTORIA      91.48167
## 4      BATH BEACH      70.34000
## 5      BAY RIDGE      68.03500
## 6      BAYSIDE      71.42111
## 7  BEDFORD PARK/NORWOOD      38.24500
## 8  BEDFORD STUYVESANT      83.24172
## 9      BELMONT      56.45000
## 10     BENSONHURST      71.70429
## 11     BERGEN BEACH      73.27000
## 12     BOERUM HILL      96.57600
## 13     BOROUGH PARK      64.10857
## 14     BRIARWOOD      75.36250
## 15     BRIGHTON BEACH      81.91429
## 16     BRONX-UNKNOWN      32.06500
## 17     BRONXDALE      28.94333
## 18     BROOKLYN HEIGHTS      114.11778
## 19     BUSH TERMINAL      60.95000
## 20     BUSHWICK      76.13500
## 21     CANARSIE      46.58000
## 22     CARROLL GARDENS      93.40556
## 23     CHELSEA      215.94932
## 24     CHINATOWN      154.17952
## 25     CITY ISLAND      40.83000
## 26     CIVIC CENTER      174.06696
## 27     CLINTON      176.70032
## 28     CLINTON HILL      88.97385
## 29     COBBLE HILL      120.69800
## 30     COBBLE HILL-WEST      85.71125
## 31     COLLEGE POINT      65.05000
## 32     CONEY ISLAND      55.05750
## 33     CORONA      94.20706
## 34     CROWN HEIGHTS      64.26286
## 35  DOWNTOWN-FULTON FERRY      103.26857
## 36  DOWNTOWN-FULTON MALL      132.42500
## 37  DOWNTOWN-METROTECH      122.48000
## 38     DYKER HEIGHTS      68.36000
## 39     EAST NEW YORK      36.99167
## 40     EAST TREMONT      72.33333
## 41     EAST VILLAGE      207.46115
## 42     ELMHURST      69.80564
## 43     FAR ROCKAWAY      74.88500
```

## 44	FASHION	194.81067
## 45	FINANCIAL	199.30917
## 46	FLATBUSH-CENTRAL	65.71167
## 47	FLATBUSH-LEFFERTS GARDEN	46.27000
## 48	FLATBUSH-NORTH	54.00000
## 49	FLATIRON	223.30311
## 50	FLUSHING-NORTH	80.16992
## 51	FLUSHING-SOUTH	89.62750
## 52	FLUSHING MEADOW PARK	58.59000
## 53	FOREST HILLS	70.20706
## 54	FORT GREENE	81.76900
## 55	GLENDALE	57.39667
## 56	GOWANUS	82.45333
## 57	GRAMERCY	188.68471
## 58	GRANT CITY	47.60000
## 59	GRAVESEND	75.63526
## 60	GREAT KILLS	33.74000
## 61	GREENPOINT	86.18053
## 62	GREENWICH VILLAGE-CENTRAL	142.57767
## 63	GREENWICH VILLAGE-WEST	202.13667
## 64	GRYMES HILL	50.09000
## 65	HAMMELS	139.07200
## 66	HARLEM-CENTRAL	102.79106
## 67	HARLEM-EAST	139.93972
## 68	HARLEM-UPPER	79.25667
## 69	HARLEM-WEST	95.20500
## 70	HIGHBRIDGE/MORRIS HEIGHTS	61.82000
## 71	HILLCREST	53.95000
## 72	HOLLIS	109.56000
## 73	HOWARD BEACH	55.06000
## 74	INWOOD	62.05500
## 75	JACKSON HEIGHTS	47.79238
## 76	JAMAICA	104.76600
## 77	JAMAICA ESTATES	79.69500
## 78	JAVITS CENTER	125.09000
## 79	KENSINGTON	56.87500
## 80	KEW GARDENS	69.64300
## 81	KINGSBRIDGE HTS/UNIV HTS	23.86000
## 82	KINGSBRIDGE/JEROME PARK	58.37800
## 83	KIPS BAY	191.31769
## 84	LITTLE ITALY	142.52308
## 85	LITTLE NECK	65.85000
## 86	LONG ISLAND CITY	108.16667
## 87	LOWER EAST SIDE	173.56262
## 88	MADISON	71.26000
## 89	MANHATTAN VALLEY	111.30043
## 90	MASPETH	53.32750
## 91	MIDDLE VILLAGE	78.35857
## 92	MIDTOWN CBD	234.36154
## 93	MIDTOWN EAST	211.04750
## 94	MIDTOWN WEST	222.06489
## 95	MIDWOOD	79.50273
## 96	MORNINGSIDE HEIGHTS	74.63000
## 97	MORRIS PARK/VAN NEST	26.90000

## 98	MORRISANIA/LONGWOOD	44.21250
## 99	MOTT HAVEN/PORT MORRIS	30.96000
## 100	MURRAY HILL	206.26795
## 101	NEW BRIGHTON	41.47667
## 102	NEW BRIGHTON-ST. GEORGE	41.06000
## 103	NEW SPRINGVILLE	40.47000
## 104	OAKLAND GARDENS	66.94000
## 105	OCEAN HILL	37.92900
## 106	OCEAN PARKWAY-NORTH	76.51111
## 107	OCEAN PARKWAY-SOUTH	75.08000
## 108	OZONE PARK	54.10000
## 109	PARK SLOPE	88.01774
## 110	PARK SLOPE SOUTH	95.84200
## 111	PARKCHESTER	32.67500
## 112	PELHAM PARKWAY SOUTH	30.55000
## 113	PROSPECT HEIGHTS	79.16200
## 114	REGO PARK	62.13630
## 115	RIDGEWOOD	64.28667
## 116	RIVERDALE	57.10176
## 117	ROCKAWAY PARK	88.13600
## 118	SCHUYLERVILLE/PELHAM BAY	49.68000
## 119	SHEEPSHEAD BAY	79.79704
## 120	SILVER LAKE	35.80500
## 121	SOHO	162.72473
## 122	SOUNDVIEW	43.40333
## 123	SOUTH OZONE PARK	40.78000
## 124	SOUTHBRIDGE	159.53333
## 125	SUNNYSIDE	61.61818
## 126	SUNSET PARK	80.58348
## 127	THROGS NECK	53.70667
## 128	TOMPKINSVILLE	35.81000
## 129	TRIBECA	180.18473
## 130	UPPER EAST SIDE (59-79)	216.83715
## 131	UPPER EAST SIDE (79-96)	202.45179
## 132	UPPER EAST SIDE (96-110)	167.41600
## 133	UPPER WEST SIDE (59-79)	200.24391
## 134	UPPER WEST SIDE (79-96)	171.84515
## 135	UPPER WEST SIDE (96-116)	134.09353
## 136	WASHINGTON HEIGHTS LOWER	65.29600
## 137	WASHINGTON HEIGHTS UPPER	93.50833
## 138	WEST NEW BRIGHTON	39.69000
## 139	WHITESTONE	72.90000
## 140	WILLIAMSBRIDGE	42.46000
## 141	WILLIAMSBURG-CENTRAL	79.97017
## 142	WILLIAMSBURG-EAST	84.32605
## 143	WILLIAMSBURG-NORTH	84.10577
## 144	WILLIAMSBURG-SOUTH	82.27618
## 145	WINDSOR TERRACE	70.21200
## 146	WOODHAVEN	38.61000
## 147	WOODSIDE	80.52625
## 148	WYCKOFF HEIGHTS	84.93000

```
print(MVNvar)
```

##	Neighborhood	Market.Value.per.SqFt
## 1	ALPHABET CITY	1.432603e+03
## 2	ARROCHAR-SHORE ACRES	NA
## 3	ASTORIA	4.765799e+02
## 4	BATH BEACH	4.725531e+02
## 5	BAY RIDGE	2.747751e+02
## 6	BAYSIDE	4.975028e+02
## 7	BEDFORD PARK/NORWOOD	1.786050e+00
## 8	BEDFORD STUYVESANT	1.686591e+02
## 9	BELMONT	NA
## 10	BENSONHURST	5.181109e+02
## 11	BERGEN BEACH	NA
## 12	BOERUM HILL	5.717304e+02
## 13	BOROUGH PARK	4.663342e+02
## 14	BRIARWOOD	7.877535e+02
## 15	BRIGHTON BEACH	1.340269e+03
## 16	BRONX-UNKNOWN	6.661250e+00
## 17	BRONXDALE	1.252563e+01
## 18	BROOKLYN HEIGHTS	1.615134e+02
## 19	BUSH TERMINAL	NA
## 20	BUSHWICK	9.316125e+01
## 21	CANARSIE	3.175682e+02
## 22	CARROLL GARDENS	3.974023e+02
## 23	CHELSEA	1.402312e+03
## 24	CHINATOWN	1.151165e+03
## 25	CITY ISLAND	NA
## 26	CIVIC CENTER	1.351097e+03
## 27	CLINTON	4.726897e+03
## 28	CLINTON HILL	1.530868e+03
## 29	COBBLE HILL	3.879921e+03
## 30	COBBLE HILL-WEST	1.500731e+02
## 31	COLLEGE POINT	6.554524e+02
## 32	CONEY ISLAND	5.252351e+02
## 33	CORONA	5.980170e+02
## 34	CROWN HEIGHTS	2.980943e+02
## 35	DOWNTOWN-FULTON FERRY	2.290867e+02
## 36	DOWNTOWN-FULTON MALL	7.916220e+02
## 37	DOWNTOWN-METROTECH	8.318775e+02
## 38	DYKER HEIGHTS	NA
## 39	EAST NEW YORK	2.959206e+02
## 40	EAST TREMONT	2.349293e+01
## 41	EAST VILLAGE	1.170189e+03
## 42	ELMHURST	7.810206e+02
## 43	FAR ROCKAWAY	1.150487e+03
## 44	FASHION	3.449067e+03
## 45	FINANCIAL	1.921830e+03
## 46	FLATBUSH-CENTRAL	6.878405e+02
## 47	FLATBUSH-LEFFERTS GARDEN	NA
## 48	FLATBUSH-NORTH	2.764193e+02
## 49	FLATIRON	1.061538e+03
## 50	FLUSHING-NORTH	6.733447e+02
## 51	FLUSHING-SOUTH	6.894765e+02
## 52	FLUSHING MEADOW PARK	NA
## 53	FOREST HILLS	6.333463e+02

## 54	FORT GREENE	3.321076e+02
## 55	GLENDAL	3.912623e+01
## 56	GOWANUS	2.689536e+02
## 57	GRAMERCY	1.562476e+03
## 58	GRANT CITY	2.784800e+02
## 59	GRAVESEND	1.678414e+02
## 60	GREAT KILLS	NA
## 61	GREENPOINT	7.792572e+01
## 62	GREENWICH VILLAGE-CENTRAL	1.450766e+03
## 63	GREENWICH VILLAGE-WEST	2.608190e+03
## 64	GRYMES HILL	NA
## 65	HAMMELS	1.267656e+04
## 66	HARLEM-CENTRAL	1.094418e+03
## 67	HARLEM-EAST	2.761448e+03
## 68	HARLEM-UPPER	3.118752e+02
## 69	HARLEM-WEST	4.500000e-04
## 70	HIGHBRIDGE/MORRIS HEIGHTS	NA
## 71	HILLCREST	NA
## 72	HOLLIS	NA
## 73	HOWARD BEACH	6.694700e+00
## 74	INWOOD	6.975113e+02
## 75	JACKSON HEIGHTS	4.740147e+01
## 76	JAMAICA	3.536060e+02
## 77	JAMAICA ESTATES	1.457460e+03
## 78	JAVITS CENTER	NA
## 79	KENSINGTON	9.112500e-01
## 80	KEW GARDENS	6.753258e+02
## 81	KINGSBRIDGE HTS/UNIV HTS	NA
## 82	KINGSBRIDGE/JEROME PARK	2.349991e+02
## 83	KIPS BAY	1.056490e+03
## 84	LITTLE ITALY	8.963593e+02
## 85	LITTLE NECK	NA
## 86	LONG ISLAND CITY	9.941855e+02
## 87	LOWER EAST SIDE	2.759558e+03
## 88	MADISON	4.268941e+02
## 89	MANHATTAN VALLEY	2.697503e+03
## 90	MASPETH	2.657642e+01
## 91	MIDDLE VILLAGE	1.217869e+03
## 92	MIDTOWN CBD	1.418938e+03
## 93	MIDTOWN EAST	1.922460e+03
## 94	MIDTOWN WEST	1.586680e+03
## 95	MIDWOOD	3.349567e+02
## 96	MORNINGSIDE HEIGHTS	NA
## 97	MORRIS PARK/VAN NEST	NA
## 98	MORRISANIA/LONGWOOD	1.801766e+02
## 99	MOTT HAVEN/PORT MORRIS	NA
## 100	MURRAY HILL	1.686214e+03
## 101	NEW BRIGHTON	1.425333e-01
## 102	NEW BRIGHTON-ST. GEORGE	5.325120e+01
## 103	NEW SPRINGVILLE	1.439005e+01
## 104	OAKLAND GARDENS	NA
## 105	OCEAN HILL	3.450968e+01
## 106	OCEAN PARKWAY-NORTH	6.847152e+02
## 107	OCEAN PARKWAY-SOUTH	NA

## 108	OZONE PARK	NA
## 109	PARK SLOPE	3.330491e+02
## 110	PARK SLOPE SOUTH	5.562082e+01
## 111	PARKCHESTER	1.185845e+01
## 112	PELHAM PARKWAY SOUTH	NA
## 113	PROSPECT HEIGHTS	7.666782e+02
## 114	REGO PARK	5.097297e+02
## 115	RIDGEWOOD	8.014532e+02
## 116	RIVERDALE	1.593113e+02
## 117	ROCKAWAY PARK	9.844819e+02
## 118	SCHUYLERVILLE/PELHAM BAY	NA
## 119	SHEEPSHEAD BAY	4.684936e+02
## 120	SILVER LAKE	5.000000e-05
## 121	SOHO	2.408419e+03
## 122	SOUNDVIEW	3.822607e+01
## 123	SOUTH OZONE PARK	NA
## 124	SOUTHBRIDGE	1.416071e+02
## 125	SUNNYSIDE	7.053084e+02
## 126	SUNSET PARK	2.718289e+02
## 127	THROGS NECK	2.675986e+02
## 128	TOMPKINSVILLE	NA
## 129	TRIBECA	2.375855e+03
## 130	UPPER EAST SIDE (59-79)	1.241306e+03
## 131	UPPER EAST SIDE (79-96)	9.892297e+02
## 132	UPPER EAST SIDE (96-110)	7.066487e+02
## 133	UPPER WEST SIDE (59-79)	2.063835e+03
## 134	UPPER WEST SIDE (79-96)	1.625411e+03
## 135	UPPER WEST SIDE (96-116)	2.660439e+03
## 136	WASHINGTON HEIGHTS LOWER	1.415999e+03
## 137	WASHINGTON HEIGHTS UPPER	8.103648e+02
## 138	WEST NEW BRIGHTON	NA
## 139	WHITESTONE	1.243010e+03
## 140	WILLIAMSBRIDGE	1.132880e+01
## 141	WILLIAMSBURG-CENTRAL	1.297635e+02
## 142	WILLIAMSBURG-EAST	1.151922e+02
## 143	WILLIAMSBURG-NORTH	2.640817e+02
## 144	WILLIAMSBURG-SOUTH	2.925146e+02
## 145	WINDSOR TERRACE	6.247337e+01
## 146	WOODHAVEN	2.566667e-02
## 147	WOODSIDE	3.862802e+02
## 148	WYCKOFF HEIGHTS	1.765279e+03

```
print(MVNsd)
```

##	Neighborhood	Market.Value.per.SqFt
## 1	ALPHABET CITY	3.784975e+01
## 2	ARROCHAR-SHORE ACRES	NA
## 3	ASTORIA	2.183071e+01
## 4	BATH BEACH	2.173829e+01
## 5	BAY RIDGE	1.657634e+01
## 6	BAYSIDE	2.230477e+01
## 7	BEDFORD PARK/NORWOOD	1.336432e+00
## 8	BEDFORD STUYVESANT	1.298688e+01
## 9	BELMONT	NA

## 10	BENSONHURST	2.276205e+01
## 11	BERGEN BEACH	NA
## 12	BOERUM HILL	2.391089e+01
## 13	BOROUGH PARK	2.159477e+01
## 14	BRIARWOOD	2.806695e+01
## 15	BRIGHTON BEACH	3.660968e+01
## 16	BRONX-UNKNOWN	2.580940e+00
## 17	BRONXDALE	3.539157e+00
## 18	BROOKLYN HEIGHTS	1.270879e+01
## 19	BUSH TERMINAL	NA
## 20	BUSHWICK	9.652008e+00
## 21	CANARSIE	1.782044e+01
## 22	CARROLL GARDENS	1.993495e+01
## 23	CHELSEA	3.744746e+01
## 24	CHINATOWN	3.392883e+01
## 25	CITY ISLAND	NA
## 26	CIVIC CENTER	3.675727e+01
## 27	CLINTON	6.875243e+01
## 28	CLINTON HILL	3.912631e+01
## 29	COBBLE HILL	6.228901e+01
## 30	COBBLE HILL-WEST	1.225043e+01
## 31	COLLEGE POINT	2.560180e+01
## 32	CONEY ISLAND	2.291801e+01
## 33	CORONA	2.445439e+01
## 34	CROWN HEIGHTS	1.726541e+01
## 35	DOWNTOWN-FULTON FERRY	1.513561e+01
## 36	DOWNTOWN-FULTON MALL	2.813578e+01
## 37	DOWNTOWN-METROTECH	2.884229e+01
## 38	DYKER HEIGHTS	NA
## 39	EAST NEW YORK	1.720234e+01
## 40	EAST TREMONT	4.846951e+00
## 41	EAST VILLAGE	3.420803e+01
## 42	ELMHURST	2.794675e+01
## 43	FAR ROCKAWAY	3.391882e+01
## 44	FASHION	5.872876e+01
## 45	FINANCIAL	4.383868e+01
## 46	FLATBUSH-CENTRAL	2.622671e+01
## 47	FLATBUSH-LEFFERTS GARDEN	NA
## 48	FLATBUSH-NORTH	1.662586e+01
## 49	FLATIRON	3.258126e+01
## 50	FLUSHING-NORTH	2.594889e+01
## 51	FLUSHING-SOUTH	2.625788e+01
## 52	FLUSHING MEADOW PARK	NA
## 53	FOREST HILLS	2.516637e+01
## 54	FORT GREENE	1.822382e+01
## 55	GLENDALE	6.255097e+00
## 56	GOWANUS	1.639981e+01
## 57	GRAMERCY	3.952817e+01
## 58	GRANT CITY	1.668772e+01
## 59	GRAVESEND	1.295536e+01
## 60	GREAT KILLS	NA
## 61	GREENPOINT	8.827554e+00
## 62	GREENWICH VILLAGE-CENTRAL	3.808892e+01
## 63	GREENWICH VILLAGE-WEST	5.107044e+01

## 64	GRYMES HILL	NA
## 65	HAMMELS	1.125902e+02
## 66	HARLEM-CENTRAL	3.308199e+01
## 67	HARLEM-EAST	5.254948e+01
## 68	HARLEM-UPPER	1.765999e+01
## 69	HARLEM-WEST	2.121320e-02
## 70	HIGHBRIDGE/MORRIS HEIGHTS	NA
## 71	HILLCREST	NA
## 72	HOLLIS	NA
## 73	HOWARD BEACH	2.587412e+00
## 74	INWOOD	2.641044e+01
## 75	JACKSON HEIGHTS	6.884872e+00
## 76	JAMAICA	1.880442e+01
## 77	JAMAICA ESTATES	3.817670e+01
## 78	JAVITS CENTER	NA
## 79	KENSINGTON	9.545942e-01
## 80	KEW GARDENS	2.598703e+01
## 81	KINGSBRIDGE HTS/UNIV HTS	NA
## 82	KINGSBRIDGE/JEROME PARK	1.532968e+01
## 83	KIPS BAY	3.250369e+01
## 84	LITTLE ITALY	2.993926e+01
## 85	LITTLE NECK	NA
## 86	LONG ISLAND CITY	3.153071e+01
## 87	LOWER EAST SIDE	5.253150e+01
## 88	MADISON	2.066142e+01
## 89	MANHATTAN VALLEY	5.193749e+01
## 90	MASPETH	5.155233e+00
## 91	MIDDLE VILLAGE	3.489799e+01
## 92	MIDTOWN CBD	3.766879e+01
## 93	MIDTOWN EAST	4.384587e+01
## 94	MIDTOWN WEST	3.983315e+01
## 95	MIDWOOD	1.830182e+01
## 96	MORNINGSIDE HEIGHTS	NA
## 97	MORRIS PARK/VAN NEST	NA
## 98	MORRISANIA/LONGWOOD	1.342299e+01
## 99	MOTT HAVEN/PORT MORRIS	NA
## 100	MURRAY HILL	4.106353e+01
## 101	NEW BRIGHTON	3.775359e-01
## 102	NEW BRIGHTON-ST. GEORGE	7.297342e+00
## 103	NEW SPRINGVILLE	3.793422e+00
## 104	OAKLAND GARDENS	NA
## 105	OCEAN HILL	5.874494e+00
## 106	OCEAN PARKWAY-NORTH	2.616706e+01
## 107	OCEAN PARKWAY-SOUTH	NA
## 108	OZONE PARK	NA
## 109	PARK SLOPE	1.824963e+01
## 110	PARK SLOPE SOUTH	7.457937e+00
## 111	PARKCHESTER	3.443610e+00
## 112	PELHAM PARKWAY SOUTH	NA
## 113	PROSPECT HEIGHTS	2.768895e+01
## 114	REGO PARK	2.257719e+01
## 115	RIDGEWOOD	2.830995e+01
## 116	RIVERDALE	1.262186e+01
## 117	ROCKAWAY PARK	3.137645e+01



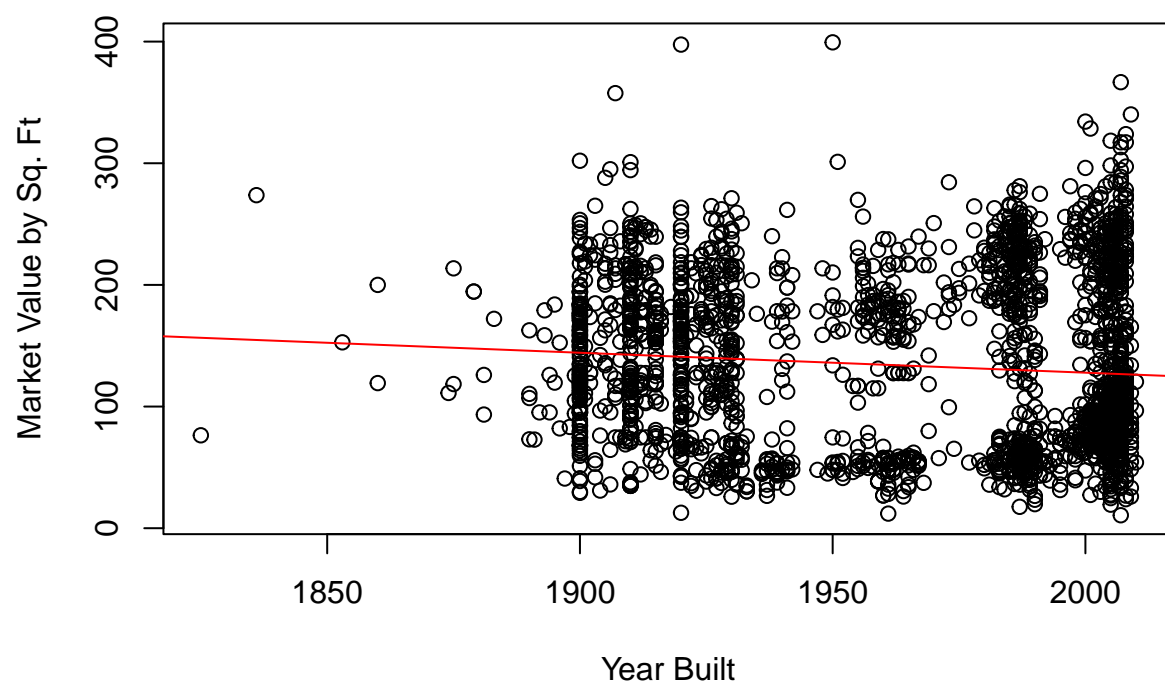
## 118	SCHUYLerville/PELHAM BAY	NA
## 119	SHEEPSHEAD BAY	2.164471e+01
## 120	SILVER LAKE	7.071068e-03
## 121	SOHO	4.907565e+01
## 122	SOUNDVIEW	6.182723e+00
## 123	SOUTH OZONE PARK	NA
## 124	SOUTHBRIDGE	1.189988e+01
## 125	SUNNYSIDE	2.655764e+01
## 126	SUNSET PARK	1.648724e+01
## 127	THROGS NECK	1.635844e+01
## 128	TOMPKINSVILLE	NA
## 129	TRIBECA	4.874274e+01
## 130	UPPER EAST SIDE (59-79)	3.523218e+01
## 131	UPPER EAST SIDE (79-96)	3.145202e+01
## 132	UPPER EAST SIDE (96-110)	2.658286e+01
## 133	UPPER WEST SIDE (59-79)	4.542946e+01
## 134	UPPER WEST SIDE (79-96)	4.031638e+01
## 135	UPPER WEST SIDE (96-116)	5.157945e+01
## 136	WASHINGTON HEIGHTS LOWER	3.762977e+01
## 137	WASHINGTON HEIGHTS UPPER	2.846691e+01
## 138	WEST NEW BRIGHTON	NA
## 139	WHITESTONE	3.525634e+01
## 140	WILLIAMSBRIDGE	3.365828e+00
## 141	WILLIAMSBURG-CENTRAL	1.139138e+01
## 142	WILLIAMSBURG-EAST	1.073276e+01
## 143	WILLIAMSBURG-NORTH	1.625059e+01
## 144	WILLIAMSBURG-SOUTH	1.710306e+01
## 145	WINDSOR TERRACE	7.904010e+00
## 146	WOODHAVEN	1.602082e-01
## 147	WOODSIDE	1.965401e+01
## 148	WYCKOFF HEIGHTS	4.201522e+01

- b. Create multiple plots to demonstrates the correlations between different variables. Remember to label all axes and give title to each graph.

```
# Enter your code here!
plot(housingData$Year.Built, housingData$Market.Value.per.SqFt,
     main = "Market Value vs. Year Built",
     xlab = "Year Built",
     ylab = "Market Value by Sq. Ft")

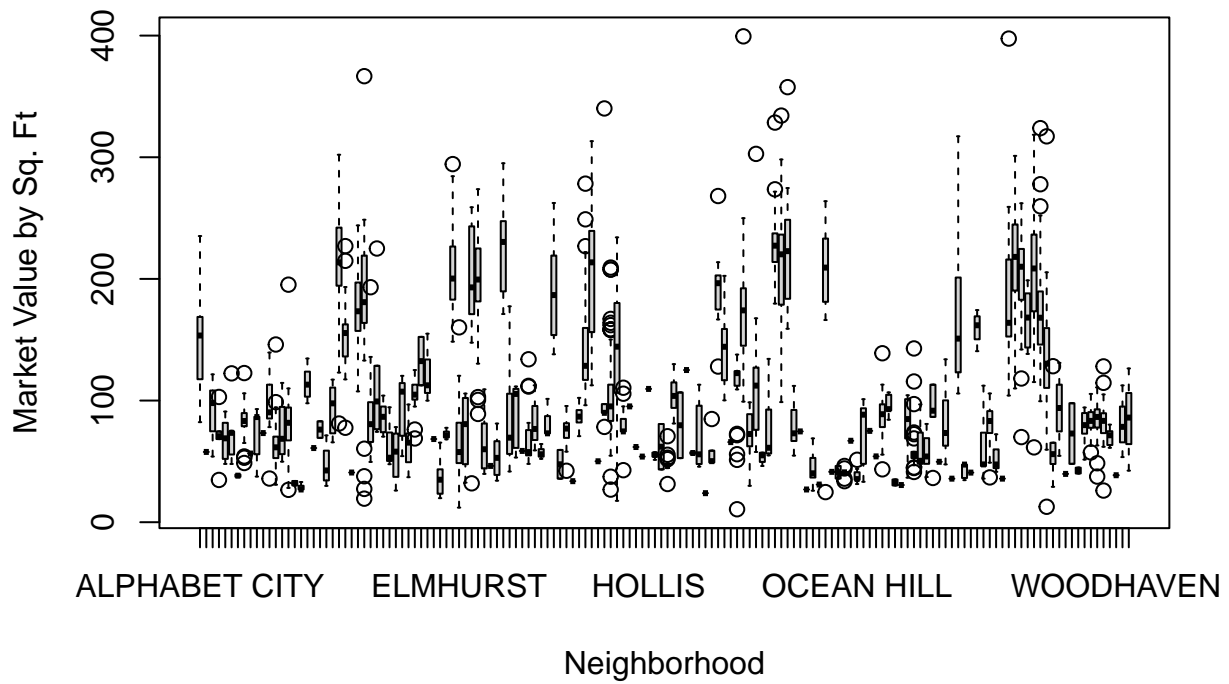
fit <- lm(Market.Value.per.SqFt ~ Year.Built, data = housingData)
abline(fit, col = "red")
```

## Market Value vs. Year Built



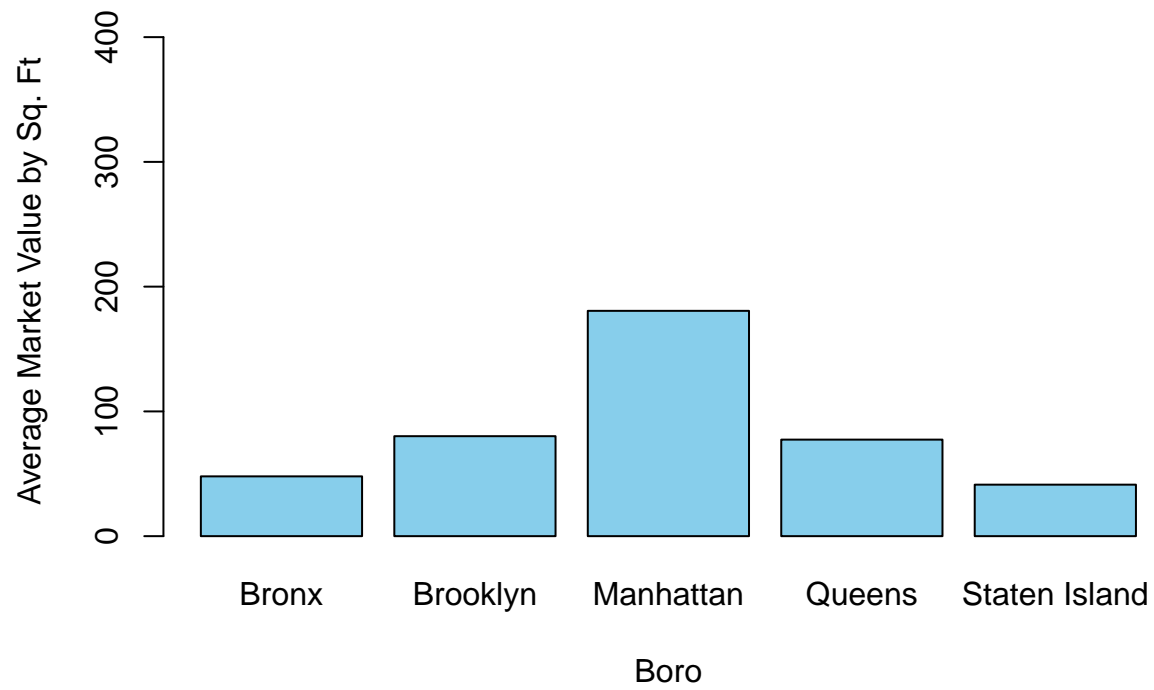
```
boxplot(Market.Value.per.SqFt ~ Neighborhood, data = housingData,  
        main = "Market Value vs. Neighborhood",  
        xlab = "Neighborhood",  
        ylab = "Market Value by Sq. Ft")
```

## Market Value vs. Neighborhood

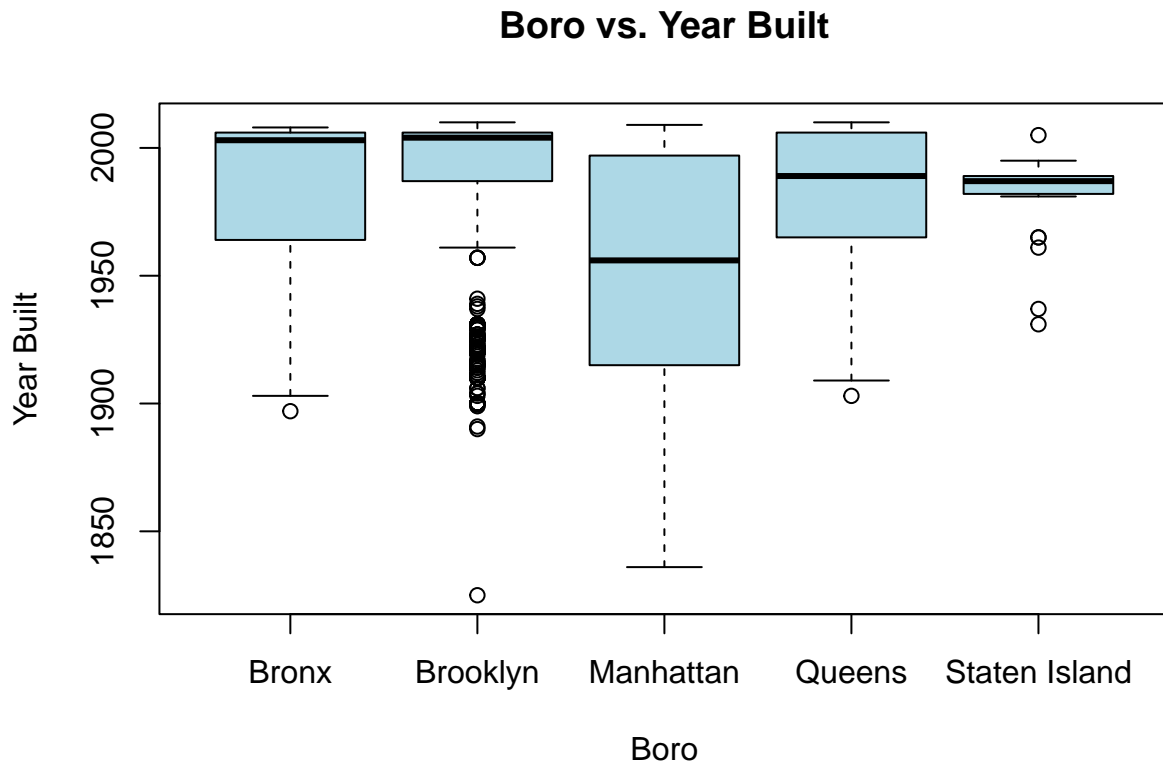


```
barplot(tapply(housingData$Market.Value.per.SqFt, housingData$Boro, mean),
  main = "Average Market Value by Boro",
  xlab = "Boro",
  ylab = "Average Market Value by Sq. Ft",
  col = "skyblue",
  ylim = c(0, max(housingData$Market.Value.per.SqFt, na.rm = TRUE) + 10))
```

## Average Market Value by Boro



```
boxplot(Year.Built ~ Boro, data = housingData,  
        main = "Boro vs. Year Built",  
        xlab = "Boro",  
        ylab = "Year Built",  
        col = "lightblue")
```



c. Write a summary about your findings from this exercise.

=> Enter your answer here!

Most of the homes were all built from 1980's to current. There aren't many buildings existing pre 1900 although there are few examples. Generally, newer houses command more market value per sq. ft despite there being many lower priced recent constructions. Manhattan commands the highest avg. market value by a decent margin followed by Brooklyn and Queens which are very similar. The Bronx is one of the cheapest boros followed by Staten Island. The surprising thing when compared to previous rankings by boro is that Manhattan is older with its median falling around 1950. Despite it's older age, it still is high priced which may speak to the neighborhood safety, location or other factors. Staten Island sticks out as well due to a limited window of development around the 1980's which is still fairly new but is the cheapest by sq. foot. This could be due to location and inconvenience associated with this location. The Bronx and Queens were both developed around same time and also are priced similarly by sq. ft. Brooklyn has been developed most recently although houses contained in dataset also reflect examples tracing back to the 1900's.