# Homework 2 Solutions

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## Part 1

i. Since NYChousing is a .csv file I use read.csv() to import the data into R.

```
setwd("~/Desktop/Data")
housing <- read.csv("NYChousing.csv", as.is = TRUE)</pre>
```

ii. The function dim() provides the dimension of its input object.

```
apply(is.na(housing), 2, sum)
```

##	UID	${\tt PropertyName}$
##	0	0
##	Lon	Lat
##	15	15
##	AgencyID	Name
##	0	0
##	Value	Address
##	52	0
##	Violations2010	REACNumber
##	0	1873
##	Borough	CD
##	0	0
##	${\tt CityCouncilDistrict}$	CensusTract
##	10	0
##	${\tt BuildingCount}$	UnitCount
##	0	0
##	YearBuilt	Owner
##	0	0
##	Rental.Coop	${\tt OwnerProfitStatus}$
##	0	0
##	${\tt AffordabilityRestrictions}$	${\tt StartAffordabilityRestrictions}$
##	0	5

The command is.na(housing) creates a matrix of the same dimensions as housing with each element being TRUE or FALSE depending on whether or not the corresponding element in housing is an NA value. Then the full call apply(is.na(housing), 2, sum) counts the number of NA values each column of housing.

iv.

```
housing <- housing[!is.na(housing$Value), ]</pre>
```

The call **is.na(housing\$Value)** returns a logical vector with TRUE where **housing\$Value** is **NA**, therefore I filter using **!is.na(housing\$Value)** to get only the rows where **Value** is not **NA**. I reassign my **housing** dataframe, to be the filtered dataframe.

v.

```
new_dim <- dim(housing)
orig_dim[1] - new_dim[1]</pre>
```

```
## [1] 52
```

I removed 52 rows of my dataframe which is what I expect, since my ouput in (iii) told me that I have 52 missing values in **Value**.

v.

```
housing$logValue <- log(housing$Value)
summary(housing$logValue)</pre>
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 8.41 12.49 13.75 13.68 14.80 20.47
```

vi.

```
housing$logUnits <- log(housing$UnitCount)</pre>
```

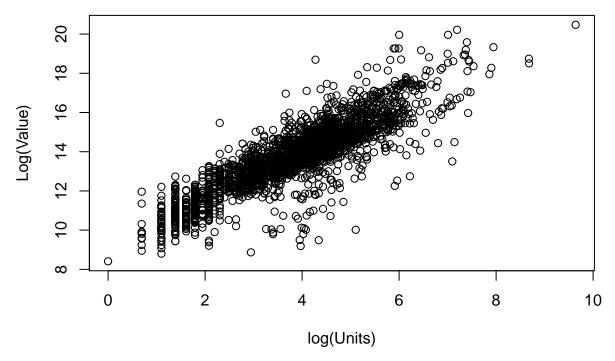
vii.

```
housing$after1950 <- housing$YearBuilt >= 1950
```

### Part 2: EDA

i.

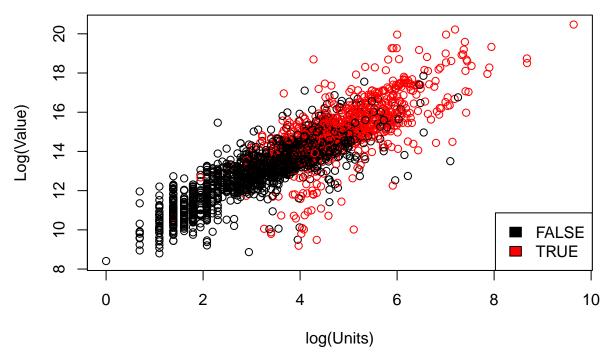
```
plot(housing$logUnits, housing$logValue, xlab = "log(Units)", ylab = "Log(Value)")
```



I plot a scatterplot with the plot() command and add argument xlab = and ylab = for the labels.

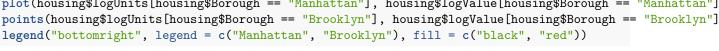
ii.

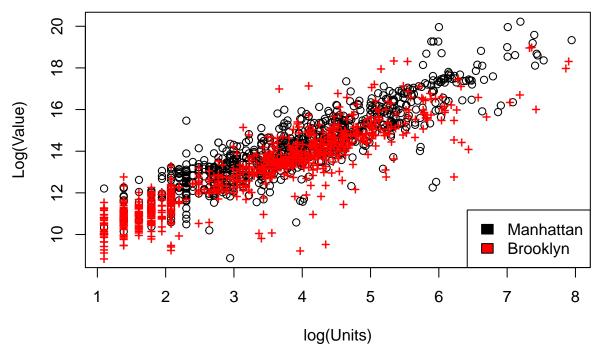
plot(housing\$logUnits, housing\$logValue, col = factor(housing\$after1950), xlab = "log(Units)", ylab = "legend("bottomright", legend = levels(factor(housing\$after1950)), fill = unique(factor(housing\$after1950))



There appears to be a pretty strong linear reltionship between logValue and logUnits. When colored according to the after1950 variable, it is clear that newer buildings (those built after 1950) tend to be more expensive and have more units than older buildings.

```
cor(housing$logValue, housing$logUnits)
## [1] 0.8727348
cor(housing$logValue[housing$Borough == "Manhattan"], housing$logUnits[housing$Borough == "Manhattan"])
## [1] 0.8830348
cor(housing$logValue[housing$Borough == "Brooklyn"], housing$logUnits[housing$Borough == "Brooklyn"])
## [1] 0.9102601
cor(housing$logValue[housing$after1950], housing$logUnits[housing$after1950])
## [1] 0.721735
cor(housing$logValue[!housing$after1950], housing$logUnits[!housing$after1950])
## [1] 0.8643297
 iv.
plot(housing$logUnits[housing$Borough == "Manhattan"], housing$logValue[housing$Borough == "Manhattan"]
```





v.

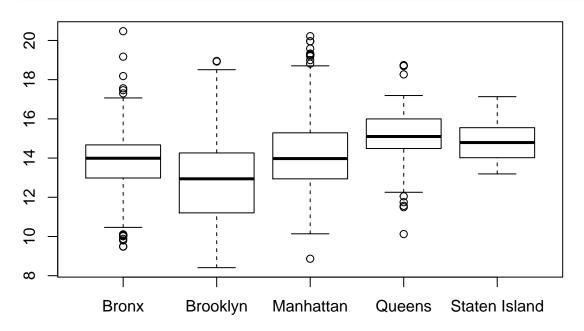
```
median(housing$Value[housing$Borough == "Manhattan"])
```

### ## [1] 1172362

The code calculates the median property value for all properties in Manhattan.

vi.

### boxplot(housing\$logValue ~ housing\$Borough)



vii.

#### tapply(housing\$Value, housing\$Borough, median)

##	Bronx	Brooklyn	Manhattan	Queens Sta	ten Island
##	1192950	417610	1172362	3611700	2654100

We use **tapply()** which splits the property value into groups based on **Borough** and then calculated the median within each group.