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HW 2 - Due Monday, Feb 19 2018 in moodle and hardcopy in class.

(1). Please upload R code and report to Moodle

with filename: HW2_IS457_YourCourseID.

(2). Turn in hard copy of your report in class.

Class ID:

In this assignment you will practice how to manipulate vector and dataframe,

such as taking subsets and creating new data structure, and end with creating a fantastic plot.

You will work with the mtcars data in R library and a dataset called SFHousing.

Before beginning with the housing data however, you will do some warm up exercises.

PART 1. Warm up (3 pts)

Q1. Create a Vector like this (0 0 2 2 4 4 6 6 8 8 10 10 12 12)

with functions seq() and rep() and call it “vec” (1 pt)

Your code below

```
x = seq(0,12,by = 2)
  vec = rep(x, each =2 )
  vec
```

```
## [1] 0 0 2 2 4 4 6 6 8 8 10 10 12 12
```

Q2. Calculate the fraction of elements in `vec` that are more than 4. (2 pts)

hint: R can do vectorized operations.

Your code below

```
sum(vec > 4)

## [1] 8

sum(vec > 4)/length(vec)

## [1] 0.5714286
```

PART II. mtcars Data (9 pts)

Q3. Use R to generate descriptions of the `mtcars` data which is already built in R base.

Print out the summary of each column and the dimensions of the dataset. (2 pts.)

(hint: you may find the `summary()` and `dim()` useful).

Write up your descriptive findings and observations of the R output. (1 pt.)

Your code below:

```
data("mtcars")
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
```

```
dim(mtcars)
```

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

Your answer below:

```
# There are 32 observations and 11 columns
```

Q4. Show last 10 cars' mpg values (1 pt.)

Your code below:

```
mtcars[23:32,c("mpg")]
```

```
## [1] 15.2 13.3 19.2 27.3 26.0 30.4 15.8 19.7 15.0 21.4
```

Q5. Show all cars' mpg values except the first 10 cars'. (1 pt.)

Your code below:

```
class(mtcars)
```

```
## [1] "data.frame"
```

```
mtcars[-(1:10),c("mpg")]
```

```
## [1] 17.8 16.4 17.3 15.2 10.4 10.4 14.7 32.4 30.4 33.9 21.5 15.5 15.2 13.3
```

```
## [15] 19.2 27.3 26.0 30.4 15.8 19.7 15.0 21.4
```

Q6. Calculate the mean of mpg subseted by "vs" variable.(1 pt)

hint: apply function family.

Your code below:

```
tapply(mtcars$mpg, mtcars$vs, mean)
```

```
##      0      1
## 16.61667 24.55714
```

Q7. Create a logical vector `mpg_vs` . (2 pts)

For the cars with V-engine (`vs = 0`), return value `TRUE` when `mpg > 14`.

For the cars with straight engine (`vs = 1`), return value `TRUE` when `mpg > 20`.

Your code below:

```
mpg_vs = c(TRUE,FALSE)

(mtcars$vs==0) & (mtcars$mpg>14)

## [1] TRUE TRUE FALSE FALSE TRUE FALSE TRUE FALSE FALSE FALSE FALSE
## [12] TRUE TRUE TRUE FALSE FALSE TRUE FALSE FALSE FALSE FALSE TRUE
## [23] TRUE FALSE TRUE FALSE TRUE FALSE TRUE TRUE TRUE FALSE

(mtcars$vs==1) & (mtcars$mpg>20)

## [1] FALSE FALSE TRUE TRUE FALSE FALSE FALSE TRUE TRUE FALSE FALSE
## [12] FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE FALSE
## [23] FALSE FALSE FALSE TRUE FALSE TRUE FALSE FALSE FALSE TRUE
```

Q8. Here is an alternative way to create the same vector in Q2.

First, we create a numeric vector `mpg_index` that is 14 for each car with V-engine

and 20 for each car with straight engine. To do this, first create a vector of length 2 called

`id_val` whose first element is 14 and second element is 20. (1 pt)

Your code below:

```
id_val = c(14,20)
          id_val

## [1] 14 20
```

Create the `mpg_index` vector by subsetting `id_val` by position, where the

positions could be represented based on `vs` column in `mtcars`. (1 pt)

Your code below

```
mpg_index= c(mtcars$vs)
mpg_index[(mpg_index==0)]= id_val[1]
mpg_index[(mpg_index==1)]= id_val[2]
mpg_index

## [1] 14 14 20 20 14 20 14 20 20 20 20 14 14 14 14 14 20 20 20 20 14 14
## [24] 14 14 20 14 20 14 14 14 20
```

Finally, use `mpg_index` and `mpg` column to create the desired vector, and

call it `mpg_vs2`. (1 pt)

Your code below

```
mpg_vs2 = data.frame(mpg_index,mtcars$mpg)
mpg_vs2

##   mpg_index mtcars.mpg
## 1         14        21.0
## 2         14        21.0
## 3         20        22.8
## 4         20        21.4
## 5         14        18.7
## 6         20        18.1
## 7         14        14.3
## 8         20        24.4
## 9         20        22.8
## 10        20        19.2
## 11        20        17.8
## 12        14        16.4
## 13        14        17.3
## 14        14        15.2
## 15        14        10.4
## 16        14        10.4
## 17        14        14.7
## 18        20        32.4
## 19        20        30.4
## 20        20        33.9
## 21        20        21.5
```

```
## 22      14      15.5
## 23      14      15.2
## 24      14      13.3
## 25      14      19.2
## 26      20      27.3
## 27      14      26.0
## 28      20      30.4
## 29      14      15.8
## 30      14      19.7
## 31      14      15.0
## 32      20      21.4
```

PART 3. San Francisco Housing Data (25 pts.)

Load the data into R.

```
load(url("https://www.stanford.edu/~vcs/StatData/SFHousing.rda"))
```

Q9. (3 pts.)

What objects are in SFHousing.rda? Give the name and class of each.

Your code below

```
objects()

## [1] "cities"    "housing"   "id_val"    "mpg_index" "mpg_vs"    "mpg_vs2"
## [7] "mtcars"    "vec"       "x"

class(cities)

## [1] "data.frame"

class(housing)

## [1] "data.frame"
```

Your answer here

```
#the objects in SFHousing.rda are "cities" and "housing"
#the class of cities is a dataframe
# the class of housing is a dataframe
```

Q10. give a summary of each object, including a summary of each variable and the dimension of the object. (4 pts)

Your code below

```
summary(cities)
```

```
##      longitude      latitude      county
## Min.   :-123.5   Min.    :37.01   Santa Clara County :30
## 1st Qu.: -122.5   1st Qu.:37.54   Contra Costa County:29
## Median :-122.3   Median :37.89   Marin County       :24
## Mean   :-122.3   Mean   :37.87   San Mateo County   :24
## 3rd Qu.: -122.0   3rd Qu.:38.09   Sonoma County      :23
## Max.    :-121.6   Max.    :38.80   Alameda County     :17
## NA's    :6       NA's    :6       (Other)            :16
## medianPrice      medianSize      numHouses      medianBR
## Min.   : 324000   Min.    : 861   Min.     : 11.0   Min.     :1.000
## 1st Qu.: 477500   1st Qu.:1322   1st Qu.: 138.5   1st Qu.:3.000
## Median : 605500   Median :1460   Median : 981.0   Median :3.000
## Mean   : 711043   Mean   :1565   Mean    :1727.0   Mean    :2.908
## 3rd Qu.: 800000   3rd Qu.:1672   3rd Qu.:2409.5   3rd Qu.:3.000
## Max.    :2200000   Max.    :3140   Max.    :14730.0   Max.    :4.000
##
```

```
summary(housing)
```

```
##      county      city      zip
## Santa Clara County :70424   Oakland   : 14730   94565 : 4595
## Alameda County     :60410   Santa Rosa : 9917   94509 : 4302
## Contra Costa County:59381   Fremont   : 9414   95123 : 4023
## Solano County       :23404   San Francisco: 8137   95687 : 3652
## San Mateo County    :22558   Evergreen   : 7947   94533 : 3472
## Sonoma County       :21676   Antioch     : 7726   (Other):261457
## (Other)             :23653   (Other)     :223635   NA's   : 5
## street      price      br      lsqft
## Length:281506   Min.   : 22000   Min.   :1.000   Min.   : 19
## Class :character 1st Qu.: 400000   1st Qu.:2.000   1st Qu.: 4000
## Mode  :character Median : 530000   Median :3.000   Median : 5760
## Mean   : 602000   Mean   :3.024   Mean   : 65939
## 3rd Qu.: 700000   3rd Qu.:4.000   3rd Qu.: 7701
## Max.    :20000000   Max.    :8.000   Max.    :418611600
## NA's      :21687
## bsqft      year      date
## Min.   : 122   Min.   : 0   Min.   :2003-04-27 02:00:00
## 1st Qu.: 1121   1st Qu.:1954   1st Qu.:2004-02-08 02:00:00
## Median : 1430   Median :1971   Median :2004-10-24 02:00:00
## Mean   : 1624   Mean   :1966   Mean   :2004-11-01 18:06:12
## 3rd Qu.: 1882   3rd Qu.:1985   3rd Qu.:2005-07-24 02:00:00
## Max.    :1868120   Max.    :3894   Max.    :2006-06-04 02:00:00
## NA's    :426     NA's    :9202
## long      lat
## Min.   :-123.6   Min.   :36.98
## 1st Qu.: -122.3   1st Qu.:37.50
```

```
## Median :-122.1    Median :37.77
## Mean  :-122.1    Mean   :37.78
## 3rd Qu.:-121.9    3rd Qu.:38.00
## Max.   :-121.5    Max.    :38.85
## NA's   :23316     NA's    :23316
##
##                                     quality
## QUALITY_ADDRESS_RANGE_INTERPOLATION :170719
## gpsvisualizer                       : 31084
## QUALITY_CITY_CENTROID                : 20473
## QUALITY_EXACT_PARCEL_CENTROID        : 17208
## QUALITY_ZIP_CODE_TABULATION_AREA_CENTROID: 14980
## (Other)                             : 3726
## NA's                                : 23316
##
##      match      wk
## Exact      :197044 Min.   :2003-04-21
## Relaxed     : 30570 1st Qu.:2004-02-01
## Relaxed; Soundex: 23338 Median :2004-10-18
## Soundex     : 2573  Mean  :2004-10-26
## 1           : 2244 3rd Qu.:2005-07-18
## (Other)     : 2421 Max.   :2006-05-29
## NA's       : 23316
```

```
dim(cities)
```

```
## [1] 163    7
```

```
dim(housing)
```

```
## [1] 281506    15
```

```
View(cities)
```

Q11. After exploring the data (maybe using the `summary()` function), describe in words the connection

between the two objects (e.g., what links them together). (2 pts)

Write your response here

the cities object holds information about different communities eg alameda, antioch etc stored in variables such as longitude,latitude, county etc. These variables together are used to describe each community

the housing object holds information about different houses using variables such as county,city,zip,street etc to describe attributes of each house

#What both objects have in common is the variable "county"

Q12. Describe in words two problems that you see with the data. (2 pts)

Write your response here

#the first problem I observed is that there are many missing values in both the "cities" and "housing" datasets.

secondly, I initially could not understand what information the housing dataset was trying to pass across(what things the variables were describing)

Q13. (2 pts.)

We will work the houses in Oakland, Sant Rosa, Campbell, and Sunnyvale only.

Subset the housing data frame so that we have only houses in these cities

and keep only the variables county, city, zip, price, br, bsqft, and year.

Call this new data frame SelectArea. This data frame should have 20706 observations

and 7 variables. (Note you may need to reformat any factor variables so that they

do not contain incorrect levels)

Your code below

```
SelectArea = (housing[housing$city == "Oakland" | housing$city == "Sant Rosa" |  
  housing$city == "Campbell" | housing$city == "Sunnyvale", c("county", "city", "zip",  
    "price", "br", "bsqft", "year")])  
  
View(SelectArea)  
summary(SelectArea)
```

```
##           county           city           zip  
## Alameda County :14729  Oakland :14730  94605 : 2084  
## Santa Clara County : 5976  Sunnyvale: 4062  95008 : 1914  
## Contra Costa County :    1  Campbell : 1914  94087 : 1787  
## Marin County      :    0  Alameda  :    0  94603 : 1552
```

```
## Napa County      :    0 Alamo      :    0 94621 : 1414
## San Francisco County:    0 Albany   :    0 94611 : 1357
## (Other)          :    0 (Other)   :    0 (Other):10598
## price            br            bsqft            year
## Min.      : 53000 Min.      :1.000 Min.      : 336 Min.      :1885
## 1st Qu.: 366000 1st Qu.:2.000 1st Qu.: 1026 1st Qu.:1924
## Median : 495000 Median :3.000 Median : 1283 Median :1947
## Mean    : 540766 Mean    :2.767 Mean    : 1457 Mean    :1948
## 3rd Qu.: 661000 3rd Qu.:3.000 3rd Qu.: 1720 3rd Qu.:1968
## Max.    :6750000 Max.    :8.000 Max.    :12582 Max.    :3880
##                                     NA's    :15      NA's    :398
```

Q14. (3 pts.)

We are interested in making plots of price and size of house, but before we do this

we will further subset the housing dataframe to remove the unusually large values.

Use the quantile function to determine the 95th percentile of price and bsqft

and eliminate all of those houses that are above either of these 95th percentiles

Call this new data frame SelectArea (replacing the old one) as well. It should

have 19064 observations.

Your code below

```
quantile(SelectArea$price,0.95)

##      95%
## 960000

quantile(SelectArea$bsqft, 0.95, na.rm=TRUE)

##      95%
## 2758.5

SelectArea = SelectArea[(SelectArea$price < 960000) & (SelectArea$bsqft < 2758.5), ]
SelectArea = SelectArea[!apply(is.na(SelectArea),1, all ), ]
View(SelectArea)
```

Q15. (2 pts.)

Create a new vector that is called `price__per__sqft` by dividing the sale price by the square footage

Add this new variable to the data frame.

Your code below

```
price_per_sqft = SelectArea$price/SelectArea$bsqft  
SelectArea= cbind(SelectArea, price_per_sqft)
```

Q16 (2 pts.)

Create a vector called `br__new`, that is the number of bedrooms in the house, except

when the number is greater than 6, set it (`br__new`) to 6.

Your code below

```
br_new = SelectArea$br  
View(br_new)  
br_new[br_new >= 6] = 6
```

Q17. (4 pts. 2 + 2 - see below)

Use the rainbow function to create a vector of 6 colors, call this vector rCols.

When you call this function, set the alpha argument to 0.25.

Create a vector called brCols where each element's value corresponds to the color in rCols

indexed by the number of bedrooms in the br_new.

For example, if the element in br_new is 3 then the color will be the third color in rCols.

(2 pts.)

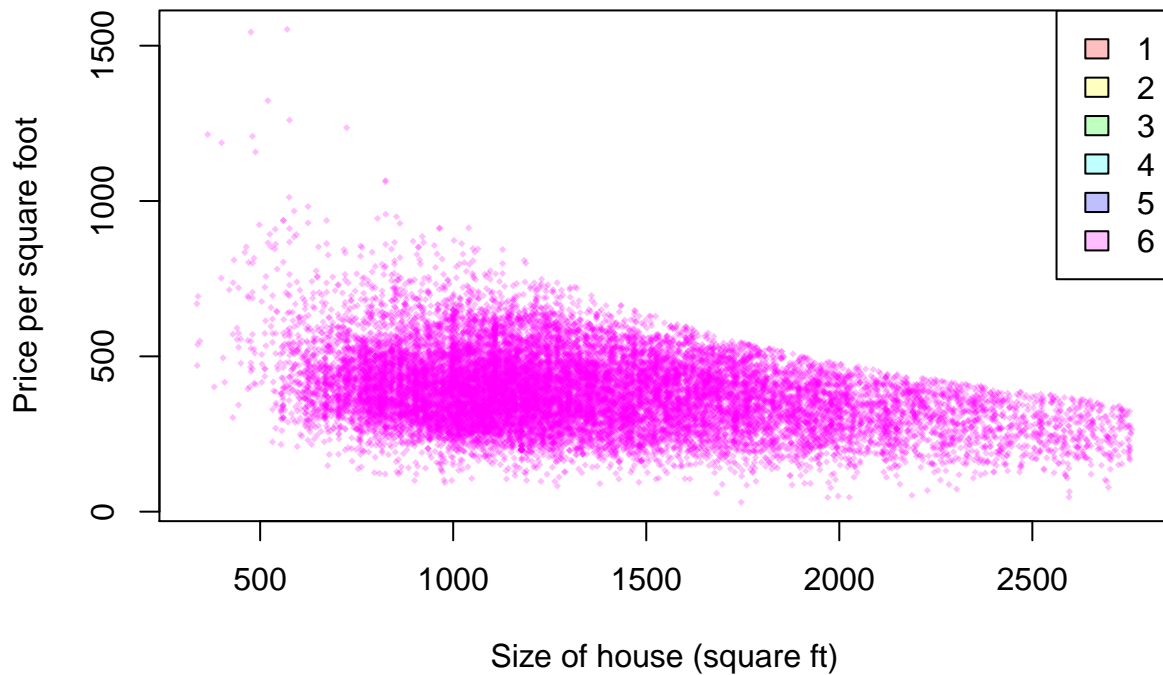
Your code below

```
rCols = rainbow(6, alpha = 0.25)
brCols = rCols[br_new]
View(brCols)
```

We are now ready to make a plot!

```
plot.new()
plot(price_per_sqft ~ bsqft, data = SelectArea,
     main = "Housing prices in the Berkeley Area",
     xlab = "Size of house (square ft)",
     ylab = "Price per square foot",
     col = brCols, pch = 18, cex = 0.5)
legend(legend = 1:6, fill = rCols, "topright")
```

Housing prices in the Berkeley Area



what's your interpretation of the plot?

e.g., the trend? the cluster? the comparison? (1 pt.)

price_per_sqft and sqft(size of the house) have a fairly neutral relationship. The increase in size of the house does not cause rise in price_per_sqft rather the smaller houses seem to have higher cost per sqft.

There also seems to be a cluster of houses bewteen 1000 - 1500 sqft.