## HW2-906338789

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#### Problem 3

I think I will definitely use it since I am always revising my codes with with new idea, so it is useful when I make a mistake and want to revert back and campare the old version with the new one. It can also be helpful by sharing it with other people so that people can work together on the same problem.

#### Problem 4

#### a. Sensory data from five operators

```
## getting "https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/Sensory.dat"
url="https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/Sensory.dat"
sensory_data_raw=fread(url,header = TRUE,fill=TRUE,skip="Item",data.table = FALSE)
saveRDS(sensory_data_raw,"sensory_data_raw.RDS")
sensory_data_raw=readRDS("sensory_data_raw.RDS")
## filling the first column with Item number
for(i in 0:9)
{
    sensory_data_raw[(3*i+2):(3*i+3),]=c(i+1,sensory_data_raw[(3*i+2):(3*i+3),])
}
```

To tidy the data, we need to push operator into a column.

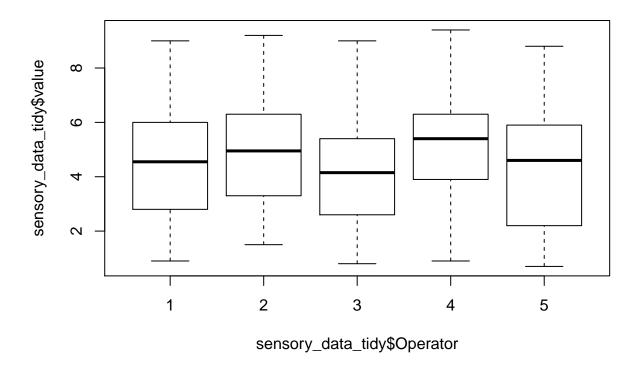
```
sensory_data_tidy=data.frame(rep(sensory_data_raw$Item,5),stack(sensory_data_raw[,-1]))
colnames(sensory_data_tidy)=c("Item","value","Operator")
head(sensory_data_tidy)
```

```
Item value Operator
##
             4.3
## 1
         1
## 2
         1
             4.3
## 3
         1
             4.1
                          1
## 4
        2
             6.0
                          1
        2
             4.9
                          1
## 5
## 6
```

We have converted the dataframes to tidy data frames using the base function. Here is a summary and boxplot of the data:

Item	value	Operator
Min.: 1.0	Min. :0.700	1:30
1st Qu.: 3.0	1st Qu.:3.025	2:30
Median: 5.5	Median $:4.700$	3:30

Item	value	Operator
Mean: 5.5	Mean :4.657	4:30
3rd Qu.: 8.0	3rd Qu.:6.000	5:30
Max.:10.0	Max. :9.400	NA



Then we choose to use tidyverse() function to tidy the raw data:

```
##
     Item operator value
                        4.3
## 1
         1
                   1
## 2
         1
                        4.3
## 3
         1
                        4.1
## 4
         2
                        6.0
## 5
         2
                   1
                        4.9
## 6
         2
                        6.0
```

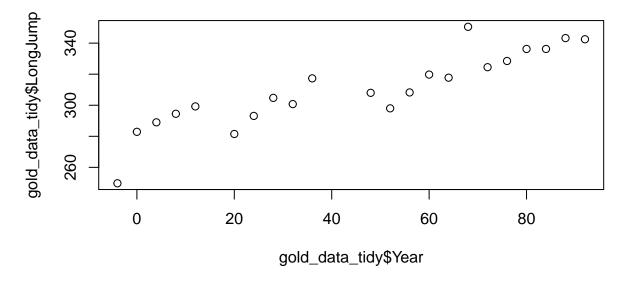
### b. Gold Medal performance for Olympic Men's Long Jump

```
## getting "https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/LongJumpData.dat"
url="https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/LongJumpData.dat"
gold_data_raw=fread(url,data.table = FALSE)
```

```
saveRDS(gold_data_raw, "gold_data_raw.RDS")
gold_data_raw=readRDS("gold_data_raw.RDS")
gold_data_raw=gold_data_raw[,1:8]
colnames(gold_data_raw)=c("Year","LongJump","Year","LongJump","Year","LongJump","Year","LongJump")
gold_data_tidy=data.frame(rbind(gold_data_raw[,1:2],gold_data_raw[,3:4]
                                 ,gold_data_raw[,5:6],gold_data_raw[,7:8]))
## Drop the raws with missing value
gold_data_tidy=DropNA(gold_data_tidy)
head(gold_data_tidy)
##
     Year LongJump
            249.75
## 1
       -4
## 2
            282.88
## 3
        4
            289.00
## 4
        8
            294.50
       12
            299.25
## 5
## 6
       20
            281.50
```

We have converted the dataframes to tidy data frames using the base function. Here is a summary and plot of the data:

Year	LongJump
Min. :-4.00	Min. :249.8
1st Qu.:21.00	1st Qu.:295.4
Median $:50.00$	Median $:308.1$
Mean $:45.45$	Mean $:310.3$
3rd Qu.:71.00	3rd Qu.:327.5
Max. $:92.00$	Max. $:350.5$



Then we choose to use tidyverse() function to tidy the raw data:

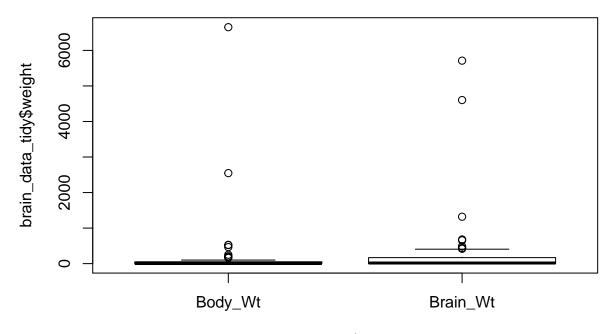
```
# stack and fix column names using tidyverse
gold_data_tv=data.frame(gather(gold_data_raw,key = "year",value="year",1,3,5,7)[,6],
                        gather(gold data raw,key="LongJump",value="LongJump",2,4,6,8)[,6])
colnames(gold data tv)=c("Year","LongJump")
head(gold data tv)
##
     Year LongJump
## 1
       -4
            249.75
            282.88
## 2
        0
## 3
        4
            289.00
## 4
       8
            294.50
            299.25
## 5
       12
## 6
       20
            281.50
```

#### c. Brain weight (g) and body weight (kg) for 62 species

```
## getting "https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/BrainandBodyWeight.dat"
url="https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/BrainandBodyWeight.dat"
brain_data_raw=fread(url,data.table = FALSE,header =TRUE,fill=TRUE)
saveRDS(brain_data_raw, "brain_data_raw.RDS")
brain_data_raw=readRDS("brain_data_raw.RDS")
colnames(brain_data_raw)=c(rep(c("Body_Wt", "Brain_Wt"),3))
brain_data_rbind=DropNA(data.frame(rbind(brain_data_raw[,1:2],
                                        brain_data_raw[,3:4],brain_data_raw[,5:6])))
brain_data_tidy=data.frame(stack(brain_data_rbind))
colnames(brain_data_tidy)=c("weight","Brain/Body")
head(brain data tidy)
##
      weight Brain/Body
## 1
       3.385
                Body Wt
## 2
       0.480
                Body_Wt
       1.350
## 3
                Body_Wt
## 4 465.000
                Body_Wt
## 5 36.330
                Body_Wt
## 6 27.660
                Body_Wt
```

We have converted the dataframes to tidy data frames using the base function. Here is a summary of the data:

weight	Brain/Body
Min. : $0.005$	Body_Wt :62
1st Qu.: 1.387	$Brain_Wt:62$
Median: 7.450	NA
Mean: $240.962$	NA
3rd Qu.: 98.650	NA
Max. :6654.000	NA



brain\_data\_tidy\$`Brain/Body`

Then we choose to use tidyverse() function to tidy the raw data:

```
# stack and fix column names using tidyverse
brain_data_tv=gather(brain_data_rbind,key="Brain/Body",value="value",Body_Wt:Brain_Wt)
head(brain_data_tv)
##
     Brain/Body
                  value
## 1
        Body_Wt
                  3.385
## 2
        Body_Wt
                  0.480
## 3
        Body_Wt
                  1.350
## 4
        Body Wt 465.000
## 5
        Body_Wt
                 36.330
## 6
        Body_Wt 27.660
```

# d. Triplicate measurements of tomato yield for two varieties of tomatos at three planting densities

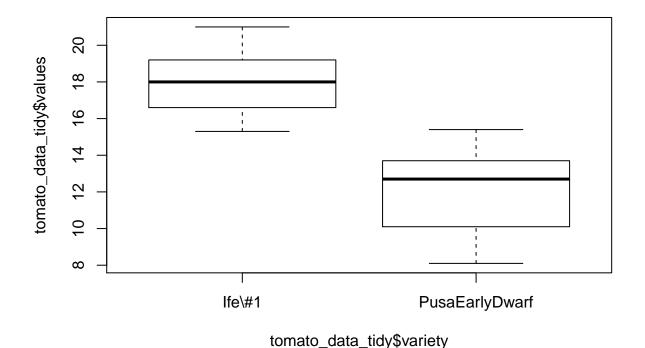
```
## getting "https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/tomato.dat"
url="https://www2.isye.gatech.edu/~jeffwu/wuhamadabook/data/tomato.dat"
tomato_data_raw=fread(url,data.table = FALSE,skip="1000")
saveRDS(tomato_data_raw,"tomato_data_raw.RDS")
tomato_data_raw=readRDS("tomato_data_raw.RDS")

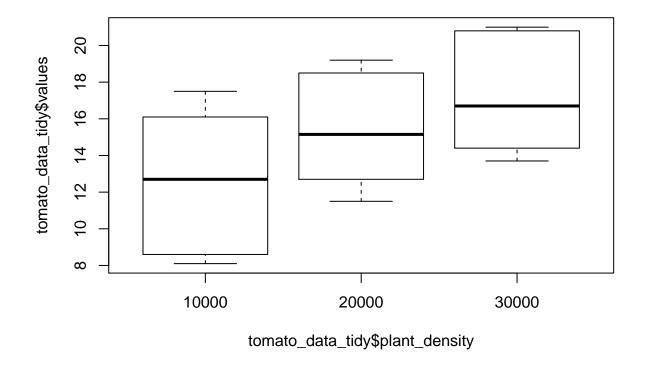
tomato_data=data.frame(varity=rep(tomato_data_raw[1:2,1],3),stack(tomato_data_raw[,-1]))
colnames(tomato_data)=c("variety","value","planting_density")
value=do.call("rbind", strsplit(tomato_data$value, ","))
value=data.frame(apply(value,2,as.numeric))
```

```
##
            variety values plant_density
## 1
                       16.1
            Ife\\#1
                                    10000
## 2 PusaEarlyDwarf
                                    10000
                        8.1
## 3
            Ife\1
                       16.6
                                    20000
## 4 PusaEarlyDwarf
                       12.7
                                    20000
            Ife\1
                       20.8
                                    30000
## 6 PusaEarlyDwarf
                       14.4
                                    30000
```

We have converted the dataframes to tidy data frames using the base function. Here is a summary and boxplots of the data:

variety	values	plant_density
Ife#1 :9	Min.: 8.10	10000:6
PusaEarlyDwarf:9	1st Qu.:12.95	20000:6
NA	Median $:15.35$	30000:6
NA	Mean : $15.07$	NA
NA	3rd Qu.:17.88	NA
NA	Max. :21.00	NA





Then we choose to use tidyverse() function to tidy the raw data:

```
# stack and fix column names using tidyverse
tomato_data_tv=gather(tomato_data_split,key="value",value="value",value1:value3)[,-3]
head(tomato_data_tv)
```

```
##
     tomato_data.variety tomato_data.planting_density value
## 1
                 Ife\1
                                                 10000
                                                        16.1
## 2
          PusaEarlyDwarf
                                                 10000
                                                         8.1
## 3
                 Ife\\#1
                                                 20000
                                                        16.6
          PusaEarlyDwarf
## 4
                                                 20000
                                                        12.7
                 Ife\\#1
## 5
                                                 30000
                                                        20.8
## 6
          PusaEarlyDwarf
                                                 30000 14.4
```