

# RWorksheet\_Camarista#4a

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1. The table below shows the data about shoe size and height. Create a data frame.

```
height <- c(66.0, 68.0, 64.5, 65.0, 70.0, 64.0, 70.0, 71.0, 72.0, 64.0, 74.5, 67.0, 71.0,
            71.0, 77.0, 72.0, 59.0, 62.0, 72.0, 66.0, 64.0, 67.0, 73.0, 69.0, 72.0, 70.0,
            69.0, 70.0)

shoeSize <- c(6.5, 9.0, 8.5, 8.5, 10.5, 7.0, 9.5, 9.0, 13.0, 7.5, 10.5, 8.5, 12.0, 10.5,
             13.0, 11.5, 8.5, 5.0, 10.0, 6.5, 7.5, 8.5, 10.5, 8.5, 10.5, 11.0, 9.0, 13.0)

shoeData <- data.frame(Shoe_Size = shoeSize, Height = height)

#got help from chat gpt to achieve the same layout from the Figure 1: Household Data,
#having 2 halves of the data frame.
half <- ceiling(nrow(shoeData) / 2)
first_half <- shoeData[1:half, ]
second_half <- shoeData[(half+1):nrow(shoeData), ]

combined <- cbind(first_half, second_half)

combined
```

##	Shoe_Size	Height	Shoe_Size	Height
## 1	6.5	66.0	13.0	77
## 2	9.0	68.0	11.5	72
## 3	8.5	64.5	8.5	59
## 4	8.5	65.0	5.0	62
## 5	10.5	70.0	10.0	72
## 6	7.0	64.0	6.5	66
## 7	9.5	70.0	7.5	64
## 8	9.0	71.0	8.5	67
## 9	13.0	72.0	10.5	73
## 10	7.5	64.0	8.5	69
## 11	10.5	74.5	10.5	72
## 12	8.5	67.0	11.0	70
## 13	12.0	71.0	9.0	69
## 14	10.5	71.0	13.0	70

```
print(str(shoeData))
```

```
## 'data.frame': 28 obs. of 2 variables:
## $ Shoe_Size: num 6.5 9 8.5 8.5 10.5 7 9.5 9 13 7.5 ...
## $ Height : num 66 68 64.5 65 70 64 70 71 72 64 ...
## NULL
```

- a. Describe the data.
  - The data is consist of 28 rows and columns(Shoe Size and Height) with numerical values.
- b. Create a subset by males and females with their corresponding shoe size and height. What its result? Show the R scripts.

```
gender <- c("F", "F", "F", "F", "M", "F", "F", "F", "M", "F", "M", "F", "M", "M", "M",
            "M", "F", "F", "M", "F", "F", "M", "M", "F", "M", "M", "M")

withGender <- cbind(shoeData, Gender = gender)

half <- ceiling(nrow(withGender) / 2)
first_half <- withGender[1:half, ]
second_half <- withGender[(half+1):nrow(withGender), ]

combinedWithGender <- cbind(first_half, second_half)

combinedWithGender
```

```
##      Shoe_Size Height Gender Shoe_Size Height Gender
## 1         6.5   66.0     F      13.0    77      M
## 2         9.0   68.0     F      11.5    72      M
## 3         8.5   64.5     F       8.5    59      F
## 4         8.5   65.0     F       5.0    62      F
## 5        10.5   70.0     M      10.0    72      M
## 6         7.0   64.0     F       6.5    66      F
## 7         9.5   70.0     F       7.5    64      F
## 8         9.0   71.0     F       8.5    67      M
## 9        13.0   72.0     M      10.5    73      M
## 10        7.5   64.0     F       8.5    69      F
## 11        10.5   74.5     M      10.5    72      M
## 12         8.5   67.0     F      11.0    70      M
## 13        12.0   71.0     M       9.0    69      M
## 14        10.5   71.0     M      13.0    70      M
```

- c. Find the mean of shoe size and height of the respondents. Write the R scripts and its result.

```
dataMeanShoeSize <- round(mean(withGender$Shoe_Size),2)
print(paste("The mean of the Shoe size is ", dataMeanShoeSize))
```

```
## [1] "The mean of the Shoe size is 9.41"
```

```
dataMeanHeight <- round(mean(withGender$Height),2)
print(paste("The mean of the Height is ", dataMeanHeight))
```

```
## [1] "The mean of the Height is 68.57"
```

- d. Is there a relationship between shoe size and height? Why?
  - “Yes, there is. As height increases, shoe size also increases. This increase in shoe or foot size is necessary to support the person’s relative height. Similarly, when constructing tall buildings, a deep foundation is required.”

**2. Construct character vector months to a factor with factor() and assign the result to factor\_months\_vector. Print out factor\_months\_vector and assert that R prints out the factor levels below the actual values.**

Consider data consisting of the names of months: “March”, “April”, “January”, “November”, “January”, “September”, “October”, “September”, “November”, “August”, “January”, “November”, “November”, “February”, “May”, “August”, “July”, “December”, “August”, “August”, “September”, “November”, “February”, “April”

```
months <- c("March", "April", "January", "November", "January", "September", "October",
            "September", "November", "August", "January", "November", "November", "February",
            "May", "August", "July", "December", "August", "August", "September", "November",
            "February", "April")

factor_months_vector <- factor(months)
factor_months_vector
```

```
## [1] March    April     January  November January  September October
## [8] September November  August   January  November November  February
## [15] May       August    July     December August   August   September
## [22] November February April
## 11 Levels: April August December February January July March May ... September
```

**3. Then check the summary() of the months\_vector and factor\_months\_vector. Interpret the results of both vectors. Are they both equally useful in this case?**

```
summary_factor_months <- summary(factor_months_vector)
summary_factor_months
```

```
##      April      August  December  February  January      July      March      May
##         2         4         1         2         3         1         1         1
## November  October  September
##         5         1         3
```

**4. Create a vector and factor for the table below.**

```

direction <- c("East", "West", "North")
frequency <- c(1, 4, 3)

factor_data <- factor(direction, levels = c("East", "West", "North"))

direction_data <- data.frame(Direction = factor_data, Frequency = frequency)

direction_data

```

```

##   Direction Frequency
## 1      East          1
## 2      West          4
## 3     North          3

```

## 5. Enter the data below in Excel with file name = import\_march.csv

-a. Import the excel file into the Environment Pane using read.table() function. Write the code.

```
import_march_data <- read.table("import_march.csv", header = TRUE, sep = ",")
```

-b. View the dataset. Write the R scripts and its result.

```
import_march_data
```

```

##   Students Strategy.1 Strategy.2 Strategy.3
## 1      Male          8          10          8
## 2              4           8           6
## 3              0           6           4
## 4     Female         14           4          15
## 5              10           2          12
## 6              6           0           9

```

## Using Conditional Statements (IF-ESE)

### 6. Fyll Search

-Exhaustive search is a methodology for finding an answer by exploring all possible cases. When trying to find a desired number in a set of given numbers, the method of finding the corresponding number by checking all elements in the set one by one can be called an exhaustive search. Implement an exhaustive search function that meets the input/output conditions below. -a. Create an R Program that allows the User to randomly select numbers from 1 to 50. Then display the chosen number. If the number is beyond the range of the selected choice, it will have to display a string "The number selected is beyond the range of 1 to 50". If number 20 is inputted by the User, it will have to display "TRUE", otherwise display the input number.

```
num_selected <- readline(prompt = "Enter a number between 1 to 50: ")
```

```
## Enter a number between 1 to 50:
```

```
if(num_selected > 50){  
    print("The number selected is beyond the range of 1 to 50")  
}else if(num_selected == 20){  
    print("TRUE")  
}else{  
    num_selected  
}
```

```
## [1] ""
```