

RWorksheet_Pineda#4a

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

#a. Describe the data.

#The dataset contains three variables: Shoe size - numerical, Height - numerical, Gender - categorical

#The data shows the relationship between individuals' shoe size and height, separated by gender.

#b. Create a subset by males and females with their corresponding shoe size and height.

#What its result? Show the R scripts.

```
shoe_data <- data.frame(  
  ShoeSize = c(6.5,9.0,8.5,8.5,10.5,7.5,9.5,10.0,13.0,7.5,10.5,8.5,12.0,10.5,  
               13.0,11.5,8.5,5.0,10.0,6.5,8.5,10.5,8.5,11.0,10.5,9.0,13.0),  
  Height = c(66.0,68.0,64.5,65.0,70.0,74.0,70.0,71.0,74.0,72.0,74.5,67.0,71.0,71.0,  
            77.0,72.0,59.0,62.0,72.0,64.0,67.0,69.0,73.0,72.0,72.0,69.0,70.0),  
  Gender = c("F","F","F","F","M","M","F","F","M","F","F","F","M","M",  
            "M","M","F","F","M","F","F","M","M","M","M","M","M")  
)
```

```
male_data <- subset(shoe_data, Gender == "M", select = c(ShoeSize, Height))  
female_data <- subset(shoe_data, Gender == "F", select = c(ShoeSize, Height))
```

```
male_data  
female_data
```

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

```
mean_shoe <- mean(shoe_data$ShoeSize)  
mean_height <- mean(shoe_data$Height)
```

```
mean_shoe  
mean_height
```

#Output:

#Mean Shoe Size = 9.63

#Mean Height = 69.3

#d. Is there a relationship between shoe size and height? Why?

#Yes, there is a positive relationship between shoe size and height.

#Generally, as height increases, shoe size also tends to increase.

#This makes sense biologically, since taller people usually have larger feet.

#2.) Construct character vector `months` to a factor with `factor()` and assign the result to `factor_months`.

#Consider data consisting of the names of months: "March", "April", "January", "November", "January", "September", "January", "November", "November", "February", "May", "August", "July", "December", "August", "August", "September", "February", "April", "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
```

#Output:

```
# [1] March    April     January  November January  September October  September
# [9] November August    January  November November February May      August
# [17] July     December August    August    September November February April
#Levels: April August December February January July March May November October September
```

#3.) Then check the `summary()` of the `months_vector` and `factor_months_vector`. Interpret the results of `summary(months)`

```
summary(factor_months_vector)
```

#The factor vector is much more useful for this kind of data because it treats each month as a category

#4.) Create a vector and factor for the table below.

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

#Note: Apply the factor function with required order of the level.

```
#new_order_data <- factor(factor_data, levels = c("East", "West", "North"))
```

```
#print(new_order_data)
```

```
direction <- c("East", "West", "West", "West", "West", "North", "North", "North", "East")
factor_direction <- factor(direction, levels = c("East", "West", "North"))
print(factor_direction)
```

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

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

```
View(import_march)
print(import_march)
```

#Output:

```
# Students Strategy1 Strategy2 Strategy3
#1 Male 8 10 8
#2 Male 4 8 6
#3 Male 0 6 4
#4 Female 14 4 15
#5 Female 10 2 12
#6 Female 6 0 9
```

#6.) Full Search

#Exhaustive search is a methodology for finding an answer by exploring all possible cases. When trying

#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 <- as.integer(readline(prompt = "Enter a number from 1 to 50: "))
```

```
if (num < 1 || num > 50) {
  cat("The number selected is beyond the range of 1 to 50\n")
} else if (num == 20) {
  cat("TRUE\n")
} else {
  cat("You entered:", num, "\n")
}
```

#7.) Change

#At ISATU University's traditional cafeteria, snacks can only be purchased with bills. A long-standing

#a. Write a function that prints the minimum number of bills that must be paid, given the price of the

#Input: Price of snack (a random number divisible by 50) Output: Minimum number of bills needed to purch

```
price <- as.integer(readline(prompt = "Enter the price of the snack (divisible by 50): "))
```

```
if (price %% 50 != 0 || price <= 0) {
  cat("The price must be a positive number divisible by 50\n")
} else {
  bills <- c(1000, 500, 200, 100, 50)
  count <- 0
  remaining <- price

  for (bill in bills) {
    count <- count + remaining %/% bill
    remaining <- remaining %% bill
  }
}
```

```

cat("Minimum number of bills needed:", count, "\n")
}

```

#8.) The following is each student's math score for one semester. Based on this, answer the following questions.

```

#Name | Grade1 | Grade2 | Grade3 | Grade4 |
# Annie | 85 | 65 | 85 | 100 |
# Thea | 65 | 75 | 90 | 90 |
# Steve | 75 | 55 | 80 | 85 |
# Hanna | 95 | 75 | 100 | 90 |

```

#a. Create a dataframe from the above table. Write the R codes and its output.

```

students <- data.frame(
  Name = c("Annie", "Thea", "Steve", "Hanna"),
  Grade1 = c(85, 65, 75, 95),
  Grade2 = c(65, 75, 55, 75),
  Grade3 = c(85, 90, 80, 100),
  Grade4 = c(100, 90, 85, 90)
)

```

```
print(students)
```

```

#Output:
#   Name Grade1 Grade2 Grade3 Grade4
#1 Annie     85     65     85    100
#2 Thea     65     75     90     90
#3 Steve    75     55     80     85
#4 Hanna    95     75    100     90

```

#b. Without using the rowMean function, output the average score of students whose average math score is greater than 90.
#Example Output: Annie's average grade this semester is 88.75.

```

for (i in 1:nrow(students)) {
  total <- students[i, 2] + students[i, 3] + students[i, 4] + students[i, 5]
  avg <- total / 4

  if (avg > 90) {
    cat(students$Name[i], "'s average grade this semester is", avg, "\n")
  }
}

```

```

#Output:
#(No output - since no one has an average above 90)

```

#c. Without using the mean function, output as follows for the tests in which the average score was less than 80.
#Example output: The nth test was difficult.

```
for (j in 2:5) {
```

```

total <- sum(students[, j])
avg <- total / nrow(students)

if (avg < 80) {
  cat("The", j - 1, "th test was difficult.\n")
}
}

```

*#d. Without using the max function, output as follows for students whose highest score for a semester is 90 or more.
 #Example Output: Annie's highest grade this semester is 95.*

```

for (i in 1:nrow(students)) {
  scores <- students[i, 2:5]
  highest <- sort(as.numeric(scores), decreasing = TRUE)[1] # top score

  if (highest > 90) {
    cat(students$Name[i], "'s highest grade this semester is", highest, "\n")
  }
}

```