# RWorksheet\_Tiad#4b

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#### 2024-10-28

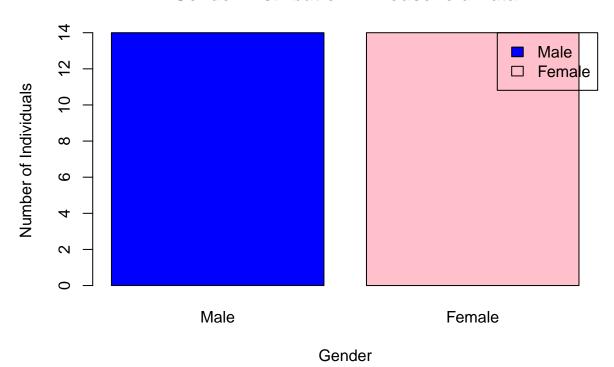
```
#USING LOOP FUNCTION
vectorA \leftarrow c(1, 2, 3, 4, 5)
matrixB <- matrix(0, nrow = 5, ncol = 5)</pre>
for (i in 1:5) {
 for (j in 1:5) {
   difference <- abs(i - j)</pre>
   matrixB[i, j] <- vectorA[difference + 1]</pre>
 }
}
print(matrixB)
       [,1] [,2] [,3] [,4] [,5]
## [1,]
        1 2
                    3
## [2,]
        2
               1
                    2
                         3
## [3,]
             2 1 2
                           3
        3
## [4,]
        4 3 2 1 2
## [5,]
for(i in 1:5){
 cat(rep("*", i), "\n")
## *
## * *
#3.
Startnum <- 10
num1 <- 0
```

```
num2 <- 1
next_num <- Startnum</pre>
repeat {
  cat(next_num, " ")
 num1 <- num2
 num2 <- next_num</pre>
 next_num <- num1 + num2</pre>
 if (next_num >= 500) {
    break
  }
}
## 10 11 21 32 53 85 138 223 361
#Using Basic Graphics (plot(),barplot(),pie(),hist())
#4.
shoes <- read.csv(file = "Shoe_Table.csv")</pre>
shoes
##
       X Shoe_size Height Gender
```

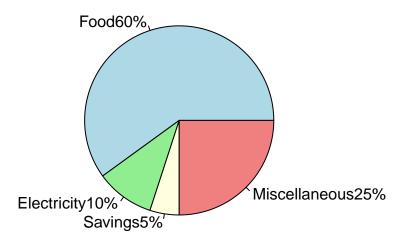
```
## 1
      1
            6.5
                  66.0
## 2 2
             9.0
                  68.0
                           F
## 3 3
             8.5
                  64.5
             8.5 65.0
                           F
## 4 4
## 5 5
            10.5
                  70.0
                           Μ
## 6 6
            7.0
                           F
                  64.0
## 7 7
            9.5
                  70.0
                           F
## 8 8
            9.0
                  71.0
                           F
## 9 9
            13.0
                  72.0
                           Μ
## 10 10
                          F
            7.5
                  64.0
## 11 11
           10.5
                  74.5
                           Μ
## 12 12
            8.5
                  67.0
                           F
## 13 13
            12.0
                  71.0
                           М
## 14 14
           10.5
                  71.0
## 15 15
           13.0
                  77.0
                           M
## 16 16
            11.5
                  72.0
                           М
## 17 17
            8.5
                  59.0
                           F
## 18 18
             5.0
                  62.0
                           F
## 19 19
            10.0
                  72.0
                           М
## 20 20
             6.5
                  66.0
                           F
                           F
## 21 21
             7.5
                  64.0
## 22 22
            8.5
                  67.0
                           Μ
## 23 23
            10.5
                  73.0
                           М
## 24 24
            8.5
                  69.0
                           F
## 25 25
            10.5
                  72.0
                           Μ
## 26 26
            11.0
                  70.0
                           M
## 27 27
             9.0
                  69.0
                           М
## 28 28
            13.0
                  70.0
                           М
```

```
head(shoes, n = 6L)
## X Shoe_size Height Gender
## 1 1
           6.5 66.0
                             F
## 2 2
            9.0 68.0
                             F
## 3 3
           8.5 64.5
                             F
           8.5
                  65.0
                             F
## 4 4
          10.5
## 5 5
                  70.0
                            Μ
           7.0 64.0
## 6 6
                             F
#b.
female_data <- subset(shoes, Gender == "F")</pre>
male_data <- subset(shoes, Gender == "M")</pre>
num_males <- nrow(male_data)</pre>
num_females <- nrow(female_data)</pre>
cat("Number of males:", num_males, "\n")
## Number of males: 14
cat("Number of females:", num_females)
## Number of females: 14
#c.
gender_counts <- c(num_males, num_females)</pre>
gender_labels <- c("Male", "Female")</pre>
barplot(gender_counts, names.arg = gender_labels, col = c("blue", "pink"),
        main = "Gender Distribution in Household Data",
        xlab = "Gender", ylab = "Number of Individuals",
        legend.text = c("Male", "Female"), args.legend = list(x = "topright"))
```

## **Gender Distribution in Household Data**



## **Dela Cruz Family Monthly Expenses**



```
#a.
data(iris)
str(iris)

## 'data.frame': 150 obs. of 5 variables:
## $ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
## $ Sepal.Width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
## $ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
## $ Petal.Width : num 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
## $ Species : Factor w/ 3 levels "setosa", "versicolor", ...: 1 1 1 1 1 1 1 1 1 1 1 ...

#b.
iris_means <- c(mean(iris$Sepal.Length), mean(iris$Sepal.Width), mean(iris$Petal.Length), mean(iris$Pet iris_means

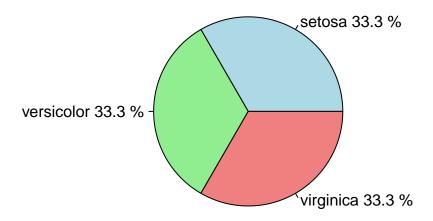
## [1] 5.843333 3.057333 3.758000 1.199333

#c.
pie(table(iris$Species),
    labels = paste(names(table(iris$Species)), round(table(iris$Species)/nrow(iris)*100, 1), "%", sep =</pre>
```

#6.

col = c("lightblue", "lightgreen", "lightcoral"),
main = "Species Distribution in Iris Dataset")

## **Species Distribution in Iris Dataset**



```
#d.
setosa <- iris[iris$Species == "setosa", ]
versicolor <- iris[iris$Species == "versicolor", ]
virginica <- iris[iris$Species == "virginica", ]
tail(setosa, 6)</pre>
```

```
Sepal.Length Sepal.Width Petal.Length Petal.Width Species
##
## 45
             5.1
                        3.8
                                   1.9
                                              0.4 setosa
             4.8
                                              0.3 setosa
## 46
                        3.0
                                    1.4
## 47
             5.1
                        3.8
                                    1.6
                                              0.2 setosa
## 48
             4.6
                        3.2
                                    1.4
                                              0.2 setosa
             5.3
## 49
                        3.7
                                    1.5
                                              0.2 setosa
                                              0.2 setosa
## 50
             5.0
                        3.3
                                    1.4
```

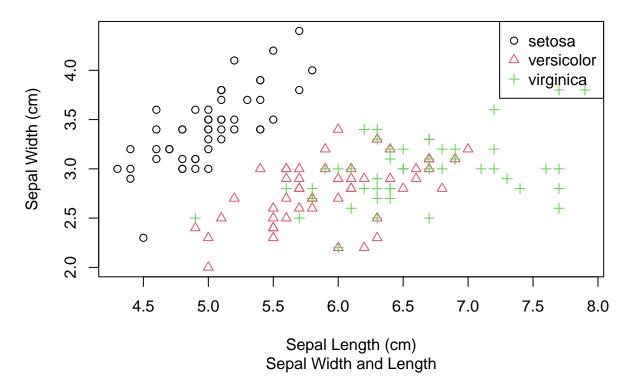
```
tail(versicolor, 6)
```

##		Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
##	95	5.6	2.7	4.2	1.3	versicolor
##	96	5.7	3.0	4.2	1.2	versicolor
##	97	5.7	2.9	4.2	1.3	versicolor
##	98	6.2	2.9	4.3	1.3	versicolor
##	99	5.1	2.5	3.0	1.1	versicolor
##	100	5.7	2.8	4.1	1.3	versicolor

#### tail(virginica, 6)

```
##
       Sepal.Length Sepal.Width Petal.Length Petal.Width
                                                               Species
## 145
                 6.7
                              3.3
                                            5.7
                                                        2.5 virginica
## 146
                 6.7
                              3.0
                                            5.2
                                                        2.3 virginica
## 147
                 6.3
                              2.5
                                            5.0
                                                        1.9 virginica
## 148
                 6.5
                              3.0
                                            5.2
                                                        2.0 virginica
## 149
                 6.2
                              3.4
                                            5.4
                                                        2.3 virginica
## 150
                 5.9
                              3.0
                                            5.1
                                                        1.8 virginica
```

#### **Iris Dataset**

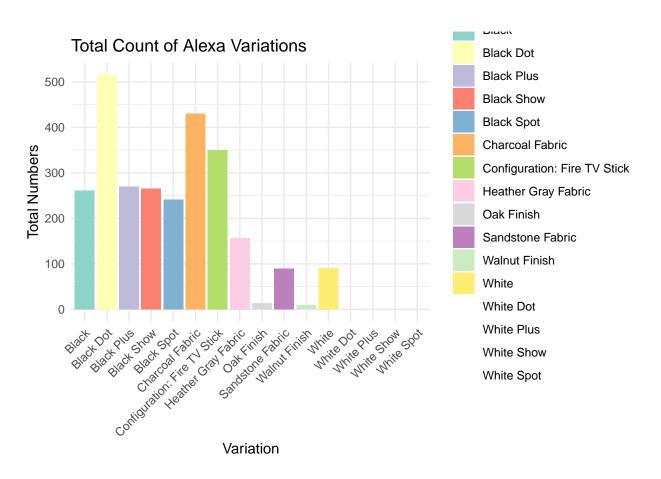


#f.
#The scatterplot shows us that the different iris species have distinct features

```
#when it comes to their sepals.
  #Setosa flowers tend to have shorter sepals and wider sepals compared to the
      #other two species.
  #Versicolor and Virginica have more similar sepal sizes, but Virginica flowers
    #generally have longer sepals than Versicolor flowers.
#This means that by looking at the sepal length and width, we can get a good
#idea of which species of iris we're dealing with!
#7.
#a.
library(readxl)
alexa_data <- read_excel("alexa_file.xlsx")</pre>
alexa_data$variation <- gsub("Black\\s+Dot", "Black Dot", alexa_data$variation)</pre>
alexa_data$variation <- gsub("Black\\s+Plus", "Black Plus", alexa_data$variation)</pre>
alexa_data$variation <- gsub("Black\\s+Show", "Black Show", alexa_data$variation)</pre>
alexa_data$variation <- gsub("Black\\s+Spot", "Black Spot", alexa_data$variation)</pre>
alexa_data$variation <- gsub("White\\s+Dot", "White Dot", alexa_data$variation)</pre>
alexa_data$variation <- gsub("White\\s+Plus", "White Plus", alexa_data$variation)
alexa_data$variation <- gsub("White\\s+Show", "White Show", alexa_data$variation)
alexa_data$variation <- gsub("White\\s+Spot", "White Spot", alexa_data$variation)
table(alexa data$variation)
##
##
                           Black
                                                     Black Dot
##
                             261
                                                            516
##
                      Black Plus
                                                    Black Show
##
                             270
                                                            265
##
                      Black Spot
                                               Charcoal Fabric
##
                                                            430
                             241
   Configuration: Fire TV Stick
                                         Heather Gray Fabric
##
                             350
                                                            157
##
                      Oak Finish
                                              Sandstone Fabric
##
                                                             90
##
                   Walnut Finish
                                                          White
                                                             91
##
                       White Dot
                                                    White Plus
##
##
                             184
                                                             78
                      White Show
                                                    White Spot
##
##
                              85
                                                            109
library(dplyr)
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
variations_count <- alexa_data %>%
  count(variation, name = "Total")
save(variations_count, file = "variations.RData")
print(variations_count)
## # A tibble: 16 x 2
##
                                   Total
     variation
##
      <chr>>
                                   <int>
## 1 Black
                                     261
## 2 Black Dot
                                     516
## 3 Black Plus
                                     270
## 4 Black Show
                                     265
## 5 Black Spot
                                     241
## 6 Charcoal Fabric
                                     430
## 7 Configuration: Fire TV Stick
                                     350
## 8 Heather Gray Fabric
                                     157
## 9 Oak Finish
                                      14
## 10 Sandstone Fabric
                                      90
## 11 Walnut Finish
                                       9
## 12 White
                                      91
## 13 White Dot
                                     184
## 14 White Plus
                                      78
## 15 White Show
                                      85
## 16 White Spot
                                     109
#c.
library(ggplot2)
load("variations.RData")
ggplot(variations_count, aes(x = variation, y = Total, fill = variation)) +
  geom_bar(stat = "identity") +
  ggtitle("Total Count of Alexa Variations") +
 xlab("Variation") +
 ylab("Total Numbers") +
 theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  scale_fill_brewer(palette = "Set3")
```

## Warning in RColorBrewer::brewer.pal(n, pal): n too large, allowed maximum for palette Set3 is 12 ## Returning the palette you asked for with that many colors



## Warning in RColorBrewer::brewer.pal(n, pal): n too large, allowed maximum for palette Set2 is 8
## Returning the palette you asked for with that many colors

### Counts of Alexa Black and White Variants

