

# worksheet#4

Dafnie Jayne Benajiba

2024-10-29

1.

```
vectorA <- c(1, 2, 3, 4, 5)

matrix_result <- matrix(0, nrow = 5, ncol = 5)

for (i in 0:4) {
  for (j in 0:4) {
    matrix_result[i + 1, j + 1] <- abs(i - j)
  }
}

print(matrix_result)
```

```
##      [,1] [,2] [,3] [,4] [,5]
## [1,]    0    1    2    3    4
## [2,]    1    0    1    2    3
## [3,]    2    1    0    1    2
## [4,]    3    2    1    0    1
## [5,]    4    3    2    1    0
```

2.

```
for (i in 1:5) {
  output <- paste(rep("'", i), collapse = " ")
  cat(output, "\n")
}
```

```
## "'"
## "'" "'"
## "'" "'" "'"
## "'" "'" "'" "'"
## "'" "'" "'" "'" "'"
```

4-A.

```
data <- read.csv("~/RBasics/CS101_DataScience/worksheet4b/Shoesize.csv")
head(data)
```

```
##   Shoe.size Height Gender Shoe.size.1 Height.1 Gender.1
## 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
```

4-B

```
males <- subset(data, Gender == 'M')
females <- subset(data, Gender == 'F')
nrow(males)
```

```
## [1] 5
```

```
nrow(females)
```

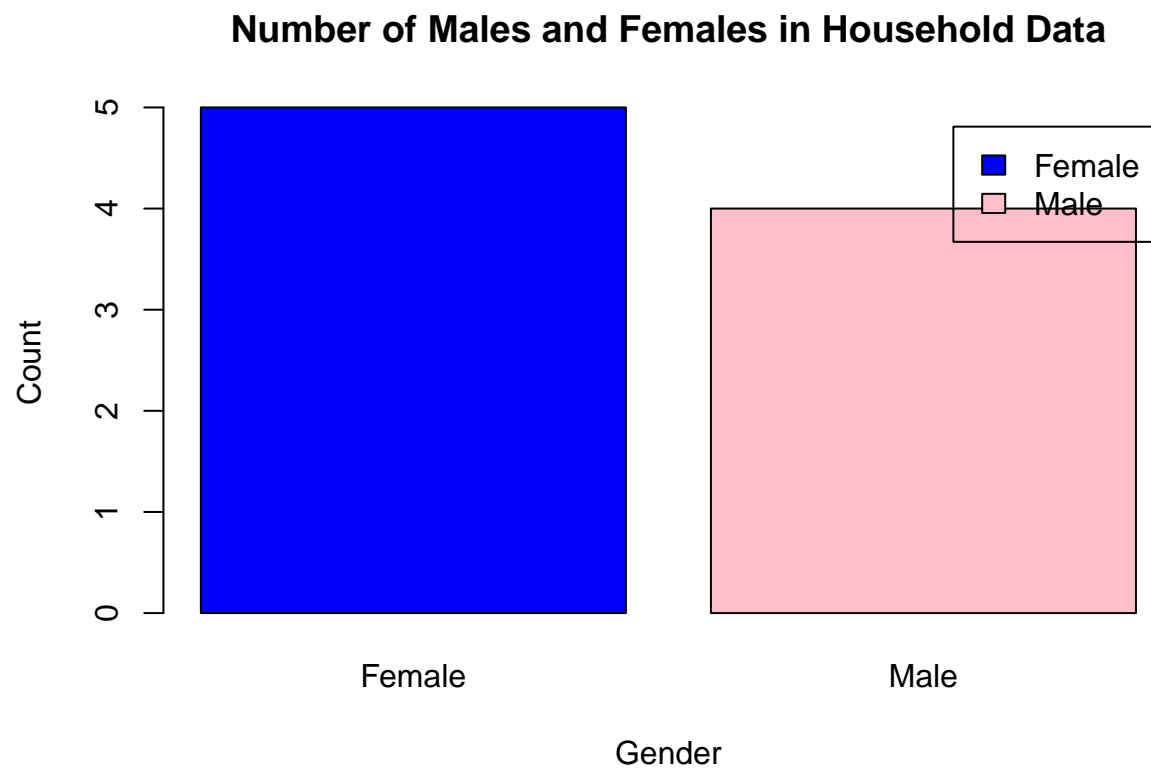
```
## [1] 9
```

4-C

```
household_data <- data.frame(
  gender = c("Male", "Female", "Male", "Female", "Male", "Female", "Male", "Female", "Female")
)

gender_counts <- table(household_data$gender)

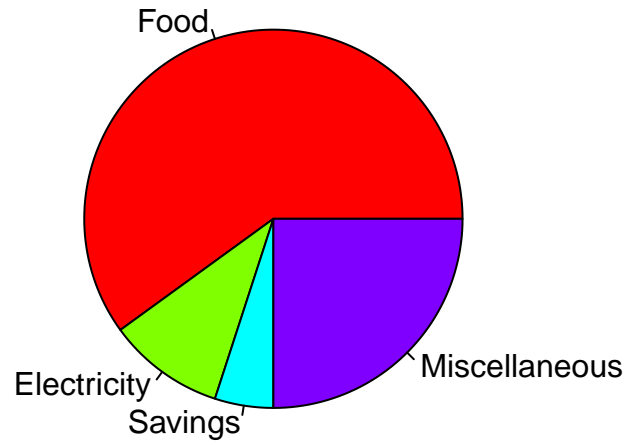
barplot(
  gender_counts,
  main = "Number of Males and Females in Household Data",
  col = c("blue", "pink"),
  legend = rownames(gender_counts),
  xlab = "Gender",
  ylab = "Count"
)
```



5.

```
expenses <- c(60, 10, 5, 25)
labels <- c("Food", "Electricity", "Savings", "Miscellaneous")
pie(expenses, labels = labels, main = "Dela Cruz Family Expenses", col=rainbow(length(expenses)))
```

## Dela Cruz Family Expenses



6-A

```
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 0.2 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 ...
```

6-B

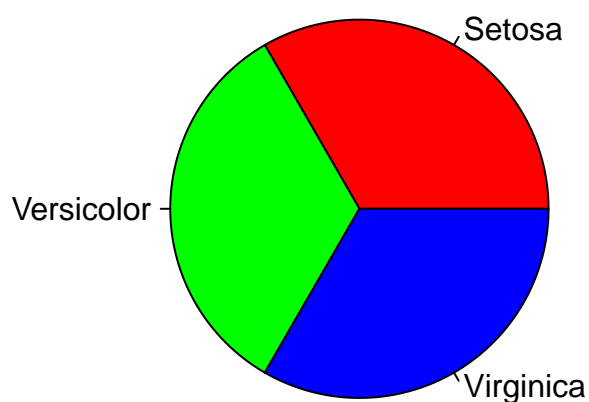
```
mean_values <- colMeans(iris[,1:4])
print(mean_values)
```

```
## Sepal.Length Sepal.Width Petal.Length Petal.Width
##      5.843333      3.057333      3.758000      1.199333
```

6-C

```
species_counts <- table(iris$Species)
pie(species_counts, main="Species Distribution", col=rainbow(3), labels=c("Setosa", "Versicolor", "Virginica"))
```

## Species Distribution



6-D

```
setosa <- subset(iris, Species == "setosa")
versicolor <- subset(iris, Species == "versicolor")
virginica <- subset(iris, Species == "virginica")
tail(setosa)
```

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

```
tail(versicolor)
```

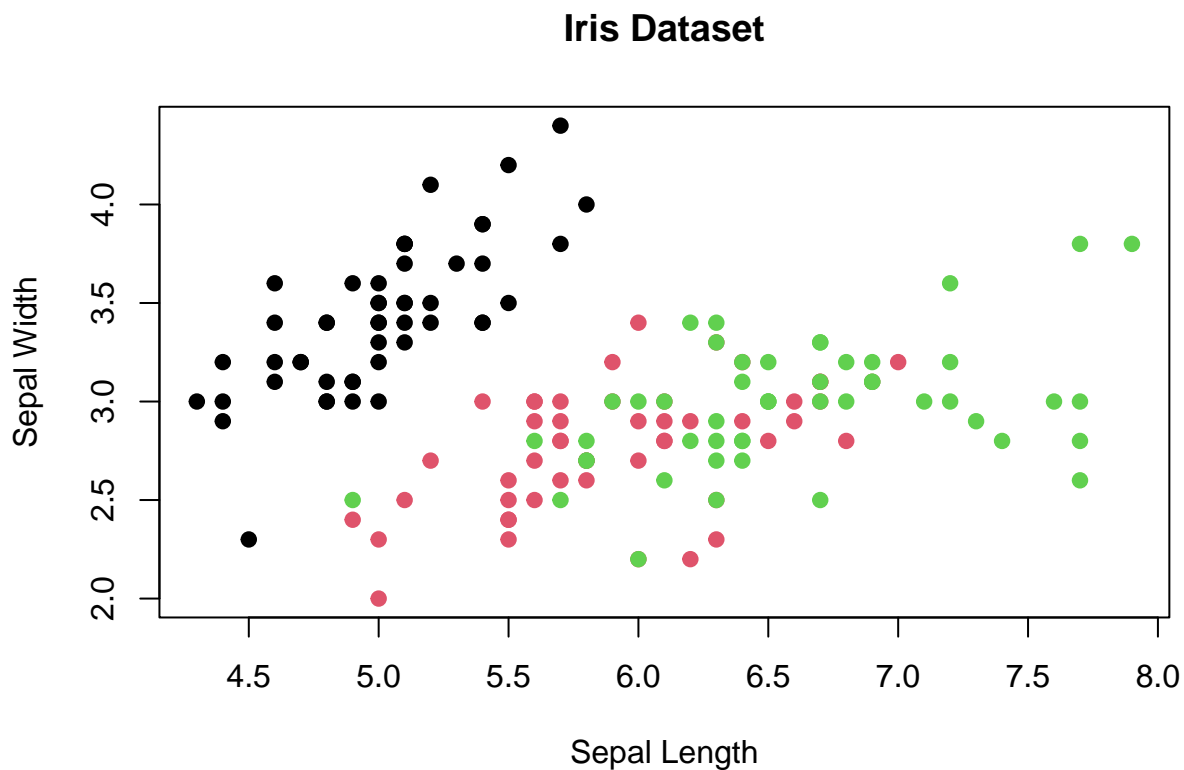
```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width  Species
## 95           5.6         2.7         4.2         1.3 versicolor
## 96           5.7         3.0         4.2         1.2 versicolor
## 97           5.7         2.9         4.2         1.3 versicolor
## 98           6.2         2.9         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)
```

```
##      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
```

6-E

```
plot(iris$Sepal.Length, iris$Sepal.Width, col=iris$Species, pch=19, main="Iris Dataset", xlab="Sepal Length", ylab="Sepal Width")
```



6-F The scatterplot shows clear separation of setosa due to its shorter sepal length and wider width, while versicolor and virginica overlap, making them harder to distinguish. Setosa is the most distinct, with more variability seen between versicolor and virginica.

7. A

```
library(readxl)
data <- read_excel("~/RBasics/CS101_DataScience/worksheet4b/alexa_file.xlsx")

data$variation <- gsub("Black Dot", "Black Dot", data$variation)
```

```
data$variation <- gsub("White Plus", "White Plus", data$variation)
```

```
head(data$variation)
```

```
## [1] "Charcoal Fabric"      "Charcoal Fabric"      "Walnut Finish"
## [4] "Charcoal Fabric"      "Charcoal Fabric"      "Heather Gray Fabric"
```

7-B

```
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
```

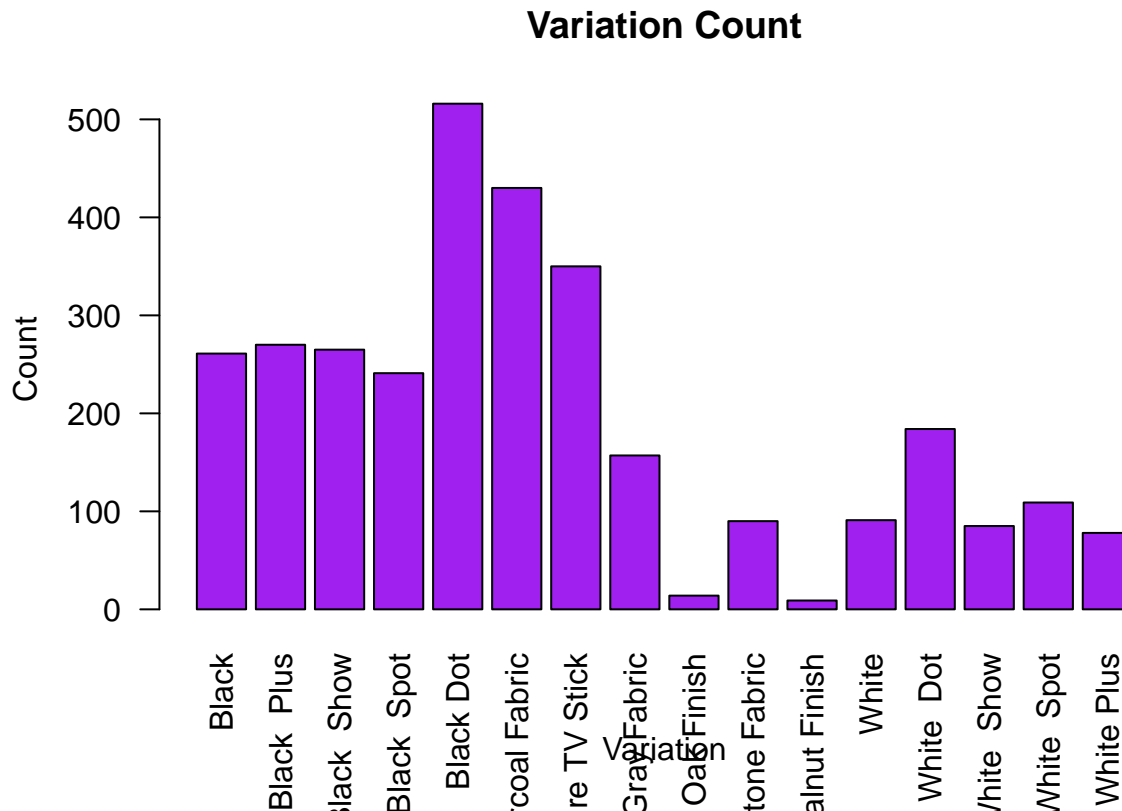
```
variation_count <- data %>%
  count(variation)
save(variation_count, file = "variations.RData")
print(variation_count)
```

```
## # A tibble: 16 x 2
##   variation      n
##   <chr>      <int>
## 1 Black      261
## 2 Black Plus 270
## 3 Black Show 265
## 4 Black Spot 241
## 5 Black Dot  516
## 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 Show  85
## 15 White Spot  109
## 16 White Plus  78
```

7b.

```
library(dplyr)
load("variations.RData")

barplot(variation_count$n, names.arg = variation_count$variation, col = "purple",
        main = "Variation Count", xlab = "Variation", ylab = "Count", las = 2)
```



7

```
library(readxl)

library(dplyr)

alexa_file <- "~/RBasics/CS101_DataScience/worksheet4b/alexa_file.xlsx"
data <- read_excel(alexa_file)

data$variation <- gsub("Black Dot", "Black Dot", data$variation)
data$variation <- gsub("Black Plus", "Black Plus", data$variation)
data$variation <- gsub("Black Show", "Black Show", data$variation)
data$variation <- gsub("Black Spot", "Black Spot", data$variation)
data$variation <- gsub("White Dot", "White Dot", data$variation)
data$variation <- gsub("White Plus", "White Plus", data$variation)
data$variation <- gsub("White Show", "White Show", data$variation)
data$variation <- gsub("White Spot", "White Spot", data$variation)

black_white_data <- data %>%
  filter(grepl("Black|White", variation))
```



```

variation_count <- black_white_data %>%
  count(variation)

black_variations <- variation_count %>% filter(grepl("Black", variation))
white_variations <- variation_count %>% filter(grepl("White", variation))

black_counts <- setNames(black_variations$n, gsub("Black ", "", black_variations$variation))
white_counts <- setNames(white_variations$n, gsub("White ", "", white_variations$variation))

common_variations <- c("Dot", "Plus", "Show", "Spot")

black_counts <- black_counts[common_variations]
black_counts[is.na(black_counts)] <- 0

white_counts <- white_counts[common_variations]
white_counts[is.na(white_counts)] <- 0

barplot_matrix <- rbind(black_counts, white_counts)

barplot(barplot_matrix, beside = TRUE, col = c("yellow", "pink"),
  main = "Count of Black and White Variations",
  xlab = "Variation Type", ylab = "Count",
  legend = rownames(barplot_matrix), args.legend = list(x = "topright"))

```

