

RWorksheet_lauron_4b.Rmd

Mary Ghale C. Lauron

2025-11-07

```
#1

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

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

for (i in 1:5) {
  for (j in 1:5) {
    matrixA[i, j] <- abs(i - j)
  }
}
```

```
# output
#print(matrixA)
#[,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) {
  for (j in 1:i) {
    cat('*')
  }
  cat("\n")
}
```

```
## "*"
## "***"
## "*****"
## "*****"
## "*****"
## "*****"
```

```
#output
#"*"
#"***"
#"*****"
#"*****"
#"*****"
#"*****"
```

```

#3 fibonacci

start <- as.numeric(readline(prompt = "Enter starting number: "))

## Enter starting number:
if (is.na(start)) {
  cat("Invalid input! Please enter a numeric value.\n")
} else {
  # Initialize first two Fibonacci numbers
  a <- 0
  b <- 1

  cat("Fibonacci sequence starting from", start, "up to 500:\n")

  repeat {
    # Only print numbers >= start
    if (a >= start) {
      cat(a, " ")
    }

    # Stop if the next number will be greater than 500
    if (a > 500) {
      break
    }

    # Generate next Fibonacci number
    temp <- a + b
    a <- b
    b <- temp
  }

  cat("\nDone!\n")
}

```

```
## Invalid input! Please enter a numeric value.
```

```
#4
```

```

#a For CSV and display first 6 rows
shoe_data <- read.csv("shoe_data.csv")

head(shoe_data)

```

```

##   Shoe_size Height Gender
## 1      6.5   66.0      F
## 2      9.0   68.0      F
## 3      8.5   64.5      F
## 4      8.5   65.0      F
## 5     10.5   70.0      M
## 6      7.0   64.0      F

```

```

#Shoe_size Height Gender
#1      6.5   66.0      F
#2      9.0   68.0      F
#3      8.5   64.5      F

```

```

#4      8.5  65.0    F
#5     10.5  70.0    M
#6      7.0  64.0    F

#b Subset for males and females
male_subset <- shoe_data[shoe_data$Gender == "M", ]
female_subset <- shoe_data[shoe_data$Gender == "F", ]
male_subset

```

```

##      Shoe_size Height Gender
## 5          10.5   70.0      M
## 9          13.0   72.0      M
## 11         10.5   74.5      M
## 13         12.0   71.0      M
## 14         10.5   71.0      M
## 15         13.0   77.0      M
## 16         11.5   72.0      M
## 19         10.0   72.0      M
## 22          8.5   67.0      M
## 23         10.5   73.0      M
## 25         10.5   72.0      M
## 26         11.0   70.0      M
## 27          9.0   69.0      M
## 28         13.0   70.0      M

```

```
female_subset
```

```

##      Shoe_size Height Gender
## 1           6.5   66.0      F
## 2           9.0   68.0      F
## 3           8.5   64.5      F
## 4           8.5   65.0      F
## 6           7.0   64.0      F
## 7           9.5   70.0      F
## 8           9.0   71.0      F
## 10          7.5   64.0      F
## 12          8.5   67.0      F
## 17          8.5   59.0      F
## 18          5.0   62.0      F
## 20          6.5   66.0      F
## 21          7.5   64.0      F
## 24          8.5   69.0      F

```

```

# Count observations in each
nrow(male_subset)

```

```
## [1] 14
```

```
nrow(female_subset)
```

```
## [1] 14
```

```

#output
#nrow(male_data)
#[1] 14
# nrow(female_data)

```

```

#[1] 14

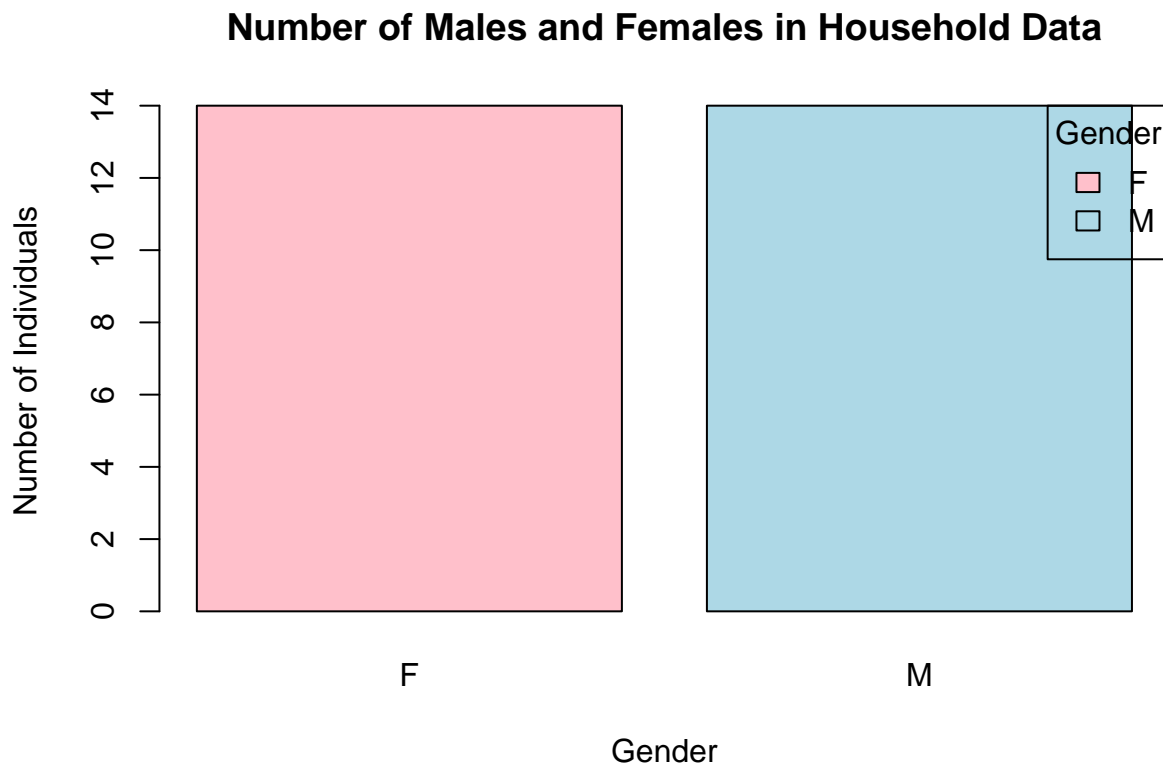
gender_count <- table(shoe_data$Gender)

# Display counts
gender_count

##
## F M
## 14 14

#c Create a bar plot
barplot(gender_count,
        main = "Number of Males and Females in Household Data",
        xlab = "Gender",
        ylab = "Number of Individuals",
        col = c("pink", "lightblue"),
        legend.text = TRUE,
        args.legend = list(x = "topright", title = "Gender"))

```



```

# 5
expenses <- c(Food = 60, Electricity = 10, Savings = 5, Miscellaneous = 25)

labels <- paste0(names(expenses),
                 " (", round(100 * expenses / sum(expenses), 1), "%)")

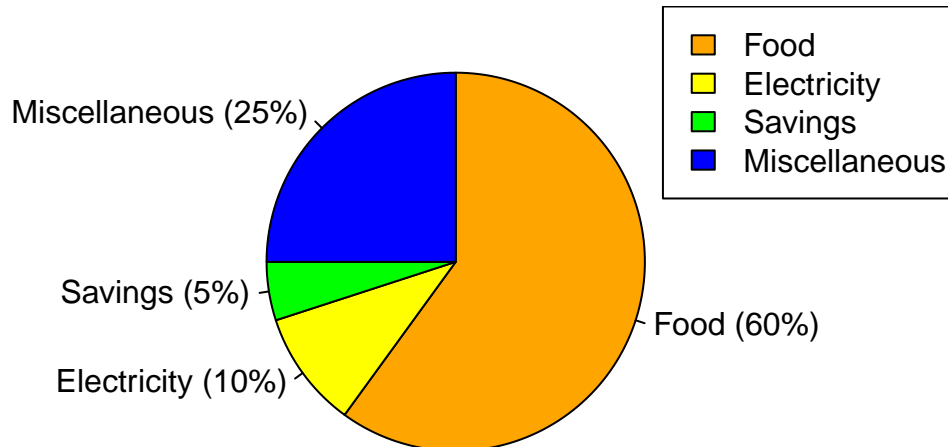
# Create pie chart
pie(expenses,
    main = "Monthly Expenses of Dela Cruz Family",

```

```
col = c("orange", "yellow", "green", "blue"),
labels = labels,
clockwise = TRUE)
```

```
legend("topright", legend = names(expenses), fill = c("orange", "yellow", "green", "blue"))
```

Monthly Expenses of Dela Cruz Family



```
#6 iris dataset
```

```
data(iris)
```

```
#a Check structure
```

```
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 ...
```

```
#output
```

```
#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 ..
```

```
# - the output shows the sepal length and sepal width, the the petal length and petal width, and the species.
```

```
#b mean of the sepal.length,sepal.width,petal.length,and petal.width.
```

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

```
iris_mean
```

```
## Sepal.Length Sepal.Width Petal.Length Petal.Width
```

```
##      5.843333      3.057333      3.758000      1.199333
```

```
#output
```

```
#Sepal.Length Sepal.Width Petal.Length Petal.Width
```

```
#      5.843333      3.057333      3.758000      1.199333

# Count species
species_distribution <- table(iris$Species)
species_distribution

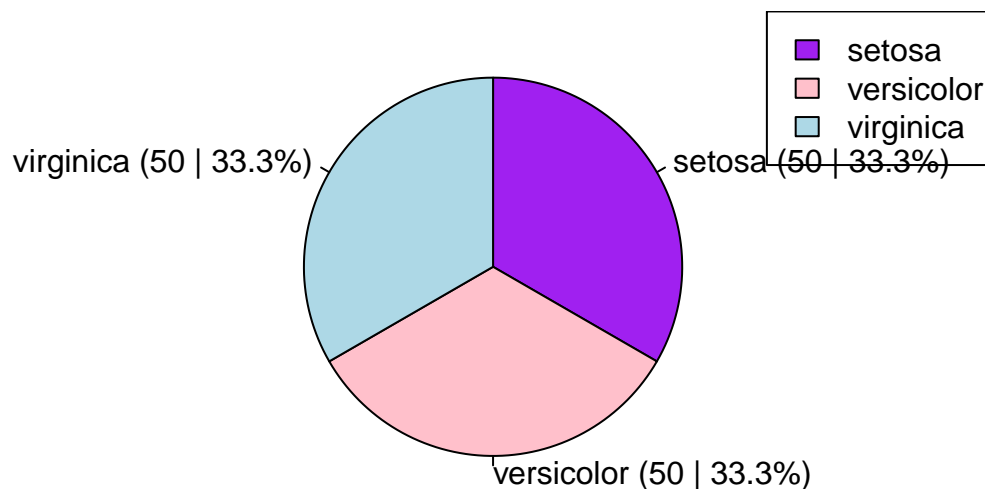
##
##      setosa versicolor  virginica
##          50         50         50

labels <- paste0(names(species_distribution),
                  " (", species_distribution, " | ",
                  round(100 * species_distribution / sum(species_distribution), 1), "%)")

# Create pie chart
pie(species_distribution,
    main = "Species Distribution in Iris Dataset",
    col = c("purple", "pink", "lightblue"),
    labels = labels,
    clockwise = TRUE)

legend("topright", legend = names(species_distribution), fill = c("purple", "pink", "lightblue"))
```

Species Distribution in Iris Dataset



```
# Subset each species
setosa <- subset(iris, Species == "setosa")
versicolor <- subset(iris, Species == "versicolor")
virginica <- subset(iris, Species == "virginica")

# last 6 rows of each
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
```

```
#output
```

```
#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
```

```
#e scatter_plot
```

```
data(iris)
```

```
iris$Species <- as.factor(iris$Species)
```

```

colors <- c("setosa" = "pink", "versicolor" = "purple", "virginica" = "blue")
symbols <- c("setosa" = 16, "versicolor" = 17, "virginica" = 18)

```

```

# Create scatterplot
plot(iris$Sepal.Length, iris$Sepal.Width,
     col = colors[iris$Species],
     pch = symbols[iris$Species],
     xlab = "Sepal Length",
     ylab = "Sepal Width",
     main = "Iris Dataset",
     sub = "Sepal width and length")

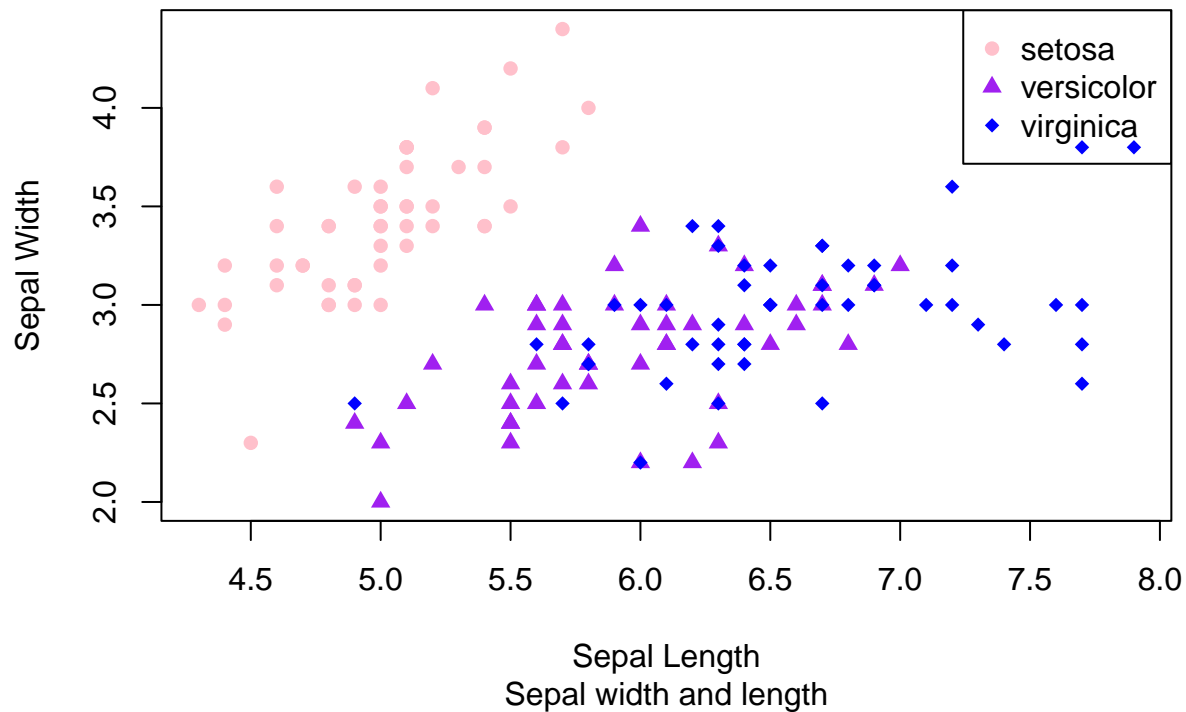
```

```

legend("topright",
      legend = levels(iris$Species),
      col = colors,
      pch = symbols)

```

Iris Dataset



#f. Interpret the result.

#Setosa species has wider sepal but shorter in length, the versicolor is in between or in average in s
#The clear separation among clusters suggests that Sepal.Length and Sepal.Width can help distinguish sp

```

#7 import alexa.xlsx

```

```

library(readxl)
alexa <- read_excel("alexa-file.xlsx")

```



```

# Show first rows
head(alexa)

## # A tibble: 6 x 5
##   rating date          variation verified_reviews feedback
##   <dbl> <dtm>          <chr>          <chr>          <dbl>
## 1     5 2018-07-30 00:00:00 Black Dot    It works great!!      1
## 2     5 2018-07-30 00:00:00 Black Dot    PHENOMENAL            1
## 3     4 2018-07-30 00:00:00 Black Dot    I used it to control my smart~ 1
## 4     5 2018-07-30 00:00:00 White Plus   Excellent device      1
## 5     3 2018-07-30 00:00:00 White Show   Very convenient      1
## 6     4 2018-07-29 00:00:00 White Dot    Great quality!        1

alexa$variation <- gsub("\\s+", " ", alexa$variation)
alexa$variation <- trimws(alexa$variation)

# Rename the variants
alexa$variation <- gsub("Black Dot", "Black - Dot", alexa$variation)
alexa$variation <- gsub("White Dot", "White - Dot", alexa$variation)
alexa$variation

## [1] "Black - Dot" "Black - Dot" "Black - Dot" "White Plus" "White Show"
## [6] "White - Dot" "Black Spot" "Black Plus" "White Plus" "White Spot"

# Output
#[1] "Black - Dot" "Black - Dot" "Black - Dot" "White Plus" "White Show" "White - Dot"
#[7] "Black Spot" "Black Plus" "White Plus" "White Spot"

# Count each variation
install.packages("dplyr")

## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.5'
## (as 'lib' is unspecified)

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 <- alexa %>%
  count(variation)

# Display result
print(variations)

## # A tibble: 7 x 2
##   variation      n
##   <chr>      <int>

```

```
## 1 Black - Dot      3
## 2 Black Plus       1
## 3 Black Spot       1
## 4 White - Dot      1
## 5 White Plus       2
## 6 White Show       1
## 7 White Spot       1

# Save the object to an RData file
save(variations, file = "variations.RData")

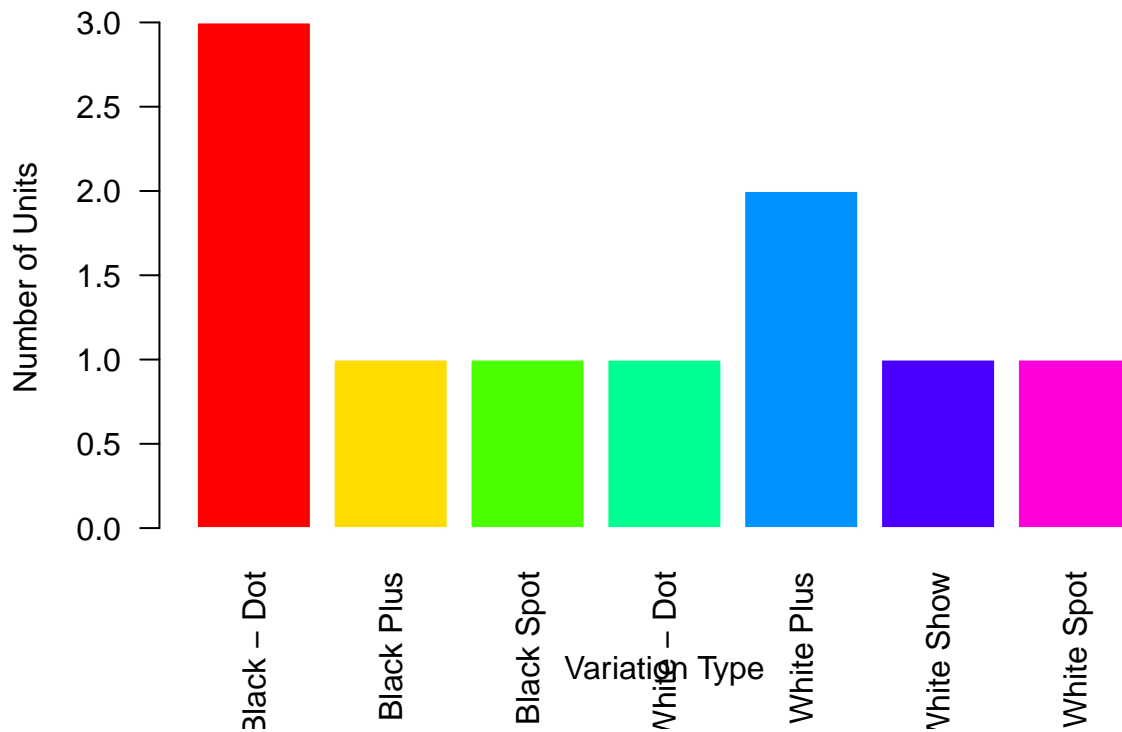
# Load the saved data (example)
load("variations.RData")
variations

## # A tibble: 7 x 2
##   variation      n
##   <chr>      <int>
## 1 Black - Dot      3
## 2 Black Plus       1
## 3 Black Spot       1
## 4 White - Dot      1
## 5 White Plus       2
## 6 White Show       1
## 7 White Spot       1

#output
#variation      n
# <chr>      <int>
#1 Black - Dot      3
#2 Black Plus       1
#3 Black Spot       1
#4 White - Dot      1
#5 White Plus       2
#6 White Show       1
#7 White Spot       1

#C
barplot(variations$n,
        names.arg = variations$variation,
        main = "Alexa Variations Distribution",
        xlab = "Variation Type",
        ylab = "Number of Units",
        col = rainbow(nrow(variations)),
        las = 2,
        border = "white")
```

Alexa Variations Distribution



```
#d

load("variations.RData")

# Separate Black and White variants
black <- variations[grep("Black", variations$variation), ]
white <- variations[grep("White", variations$variation), ]

par(mfrow = c(1, 2), mar = c(5, 5, 4, 2))

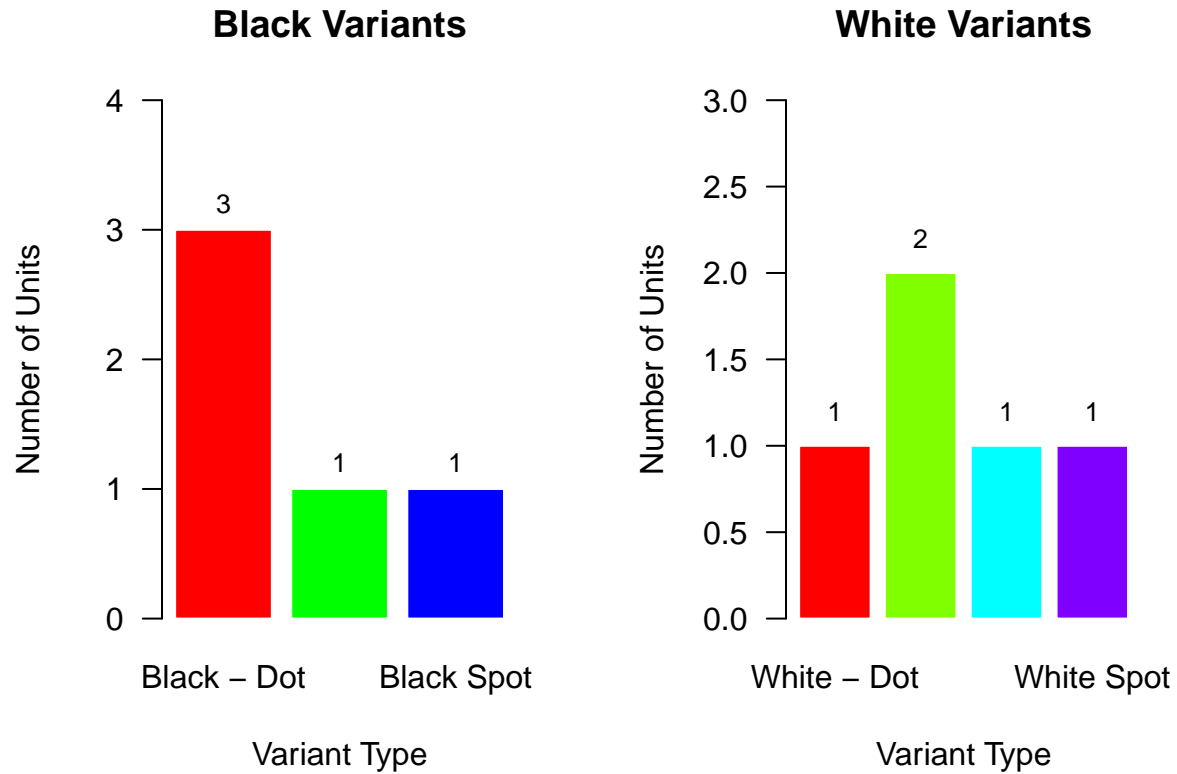
#black
bar_black <- barplot(black$n,
  names.arg = black$variation,
  main = "Black Variants",
  xlab = "Variant Type",
  ylab = "Number of Units",
  col = rainbow(nrow(black)),
  ylim = c(0, max(black$n) + 1),
  border = "white",
  las = 1)

text(x = bar_black, y = black$n + 0.2, labels = black$n, cex = 0.8, col = "black")

#white
bar_white <- barplot(white$n,
  names.arg = white$variation,
  main = "White Variants",
  xlab = "Variant Type",
  ylab = "Number of Units",
```

```
col = rainbow(nrow(white)),
ylim = c(0, max(white$n) + 1),
border = "white",
las = 1)
```

```
text(x = bar_white, y = white$n + 0.2, labels = white$n, cex = 0.8, col = "black")
```



```
par(mfrow = c(1, 1))
```