

RWorksheet_lauron_4b.Rmd

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```
#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

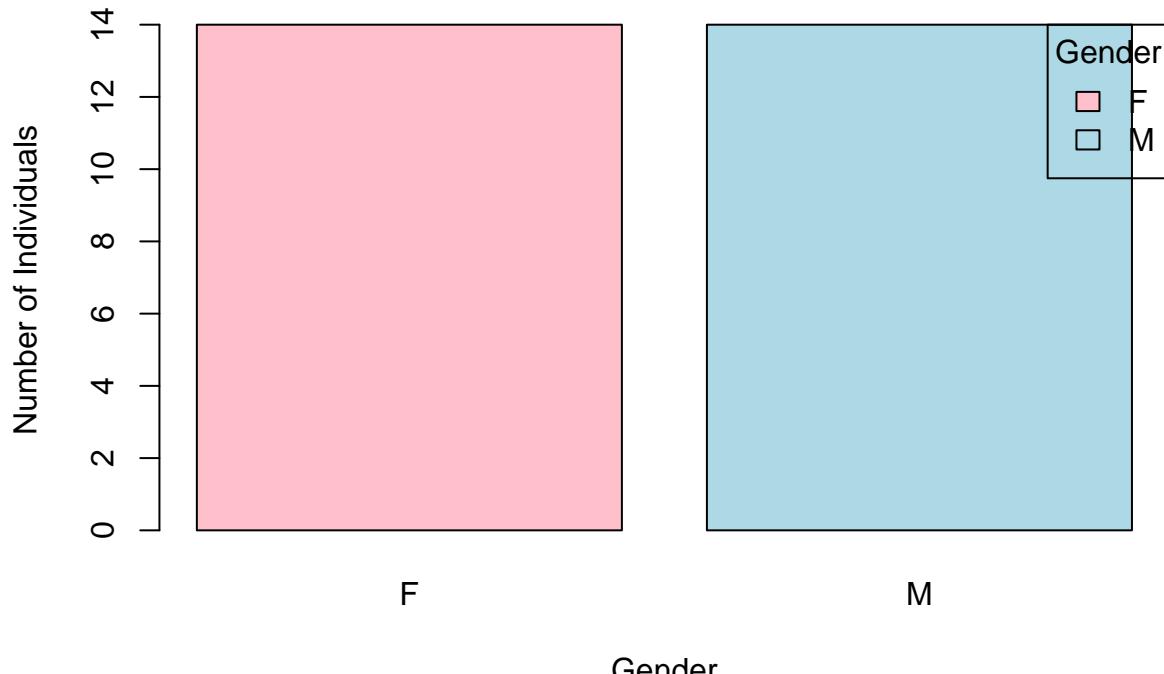
gender_count <- table(shoe_data$Gender)

# Display counts
gender_count

## 
##   F   M
## 14 14

# 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"))
```

Number of Males and Females in Household Data



```
# 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",
```

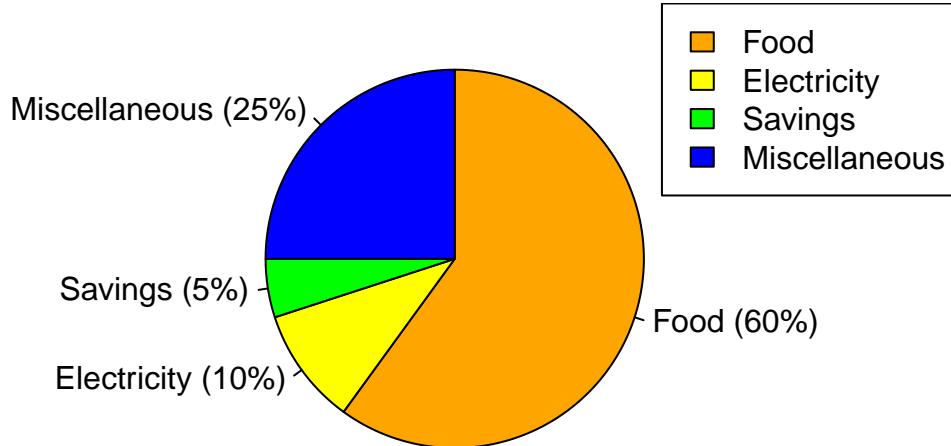
```

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 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

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

## setosa versicolor virginica
##      50          50          50

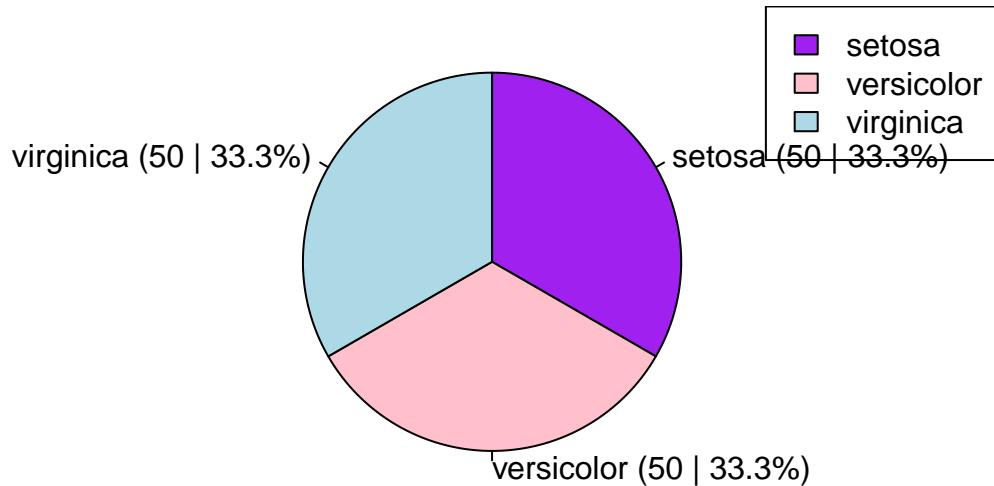
labels <- paste0(names(species_distribution),
                 " (", species_distribution, " | ",
                 round(100 * species_distribution / sum(species_distribution), 1), "%)")

# 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



```

#d 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)

```

```

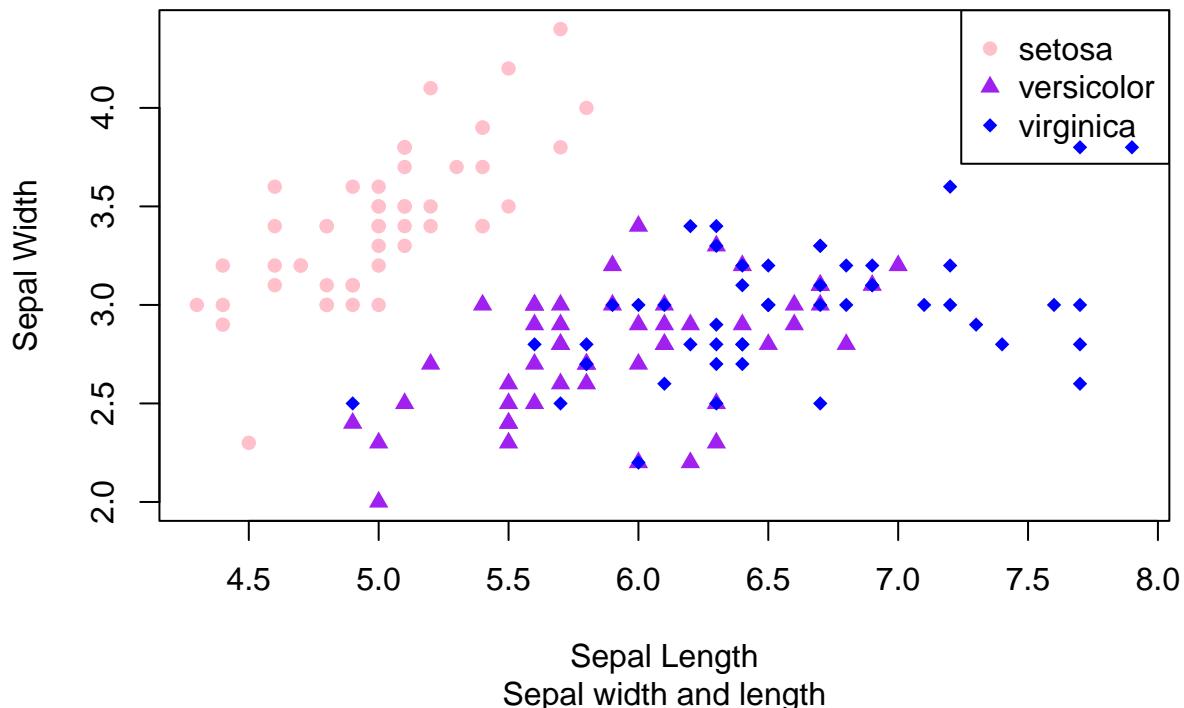
colors <- c("setosa" = "pink", "versicolor" = "purple", "virginica" = "blue")
symbols <- c("setosa" = 16, "versicolor" = 17, "virginica" = 18)

# 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> <dttm>        <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)

#a 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

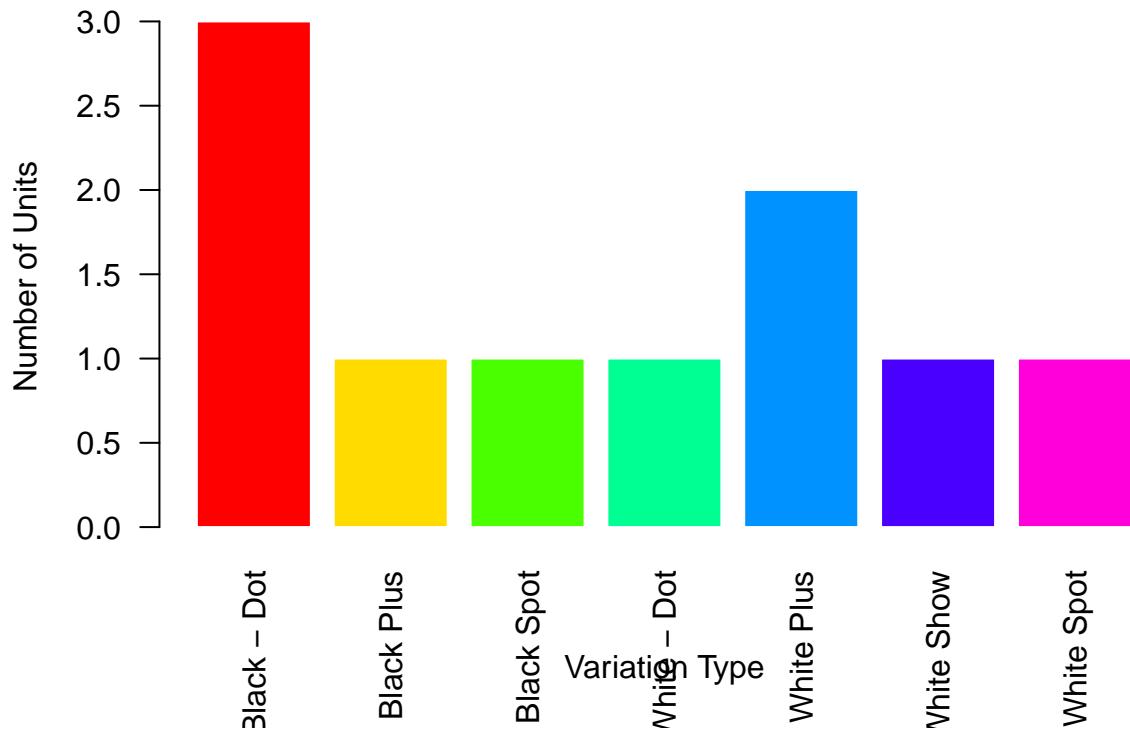
## # A tibble: 7 x 2
##   variation     n
##   <chr>     <int>
## 1 Black - Dot     3
## 2 Black Plus      1
## 3 Black Spot       1
## 4 White - Dot      1
## 5 White Plus       2
## 6 White Show        1
## 7 White Spot        1

#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)

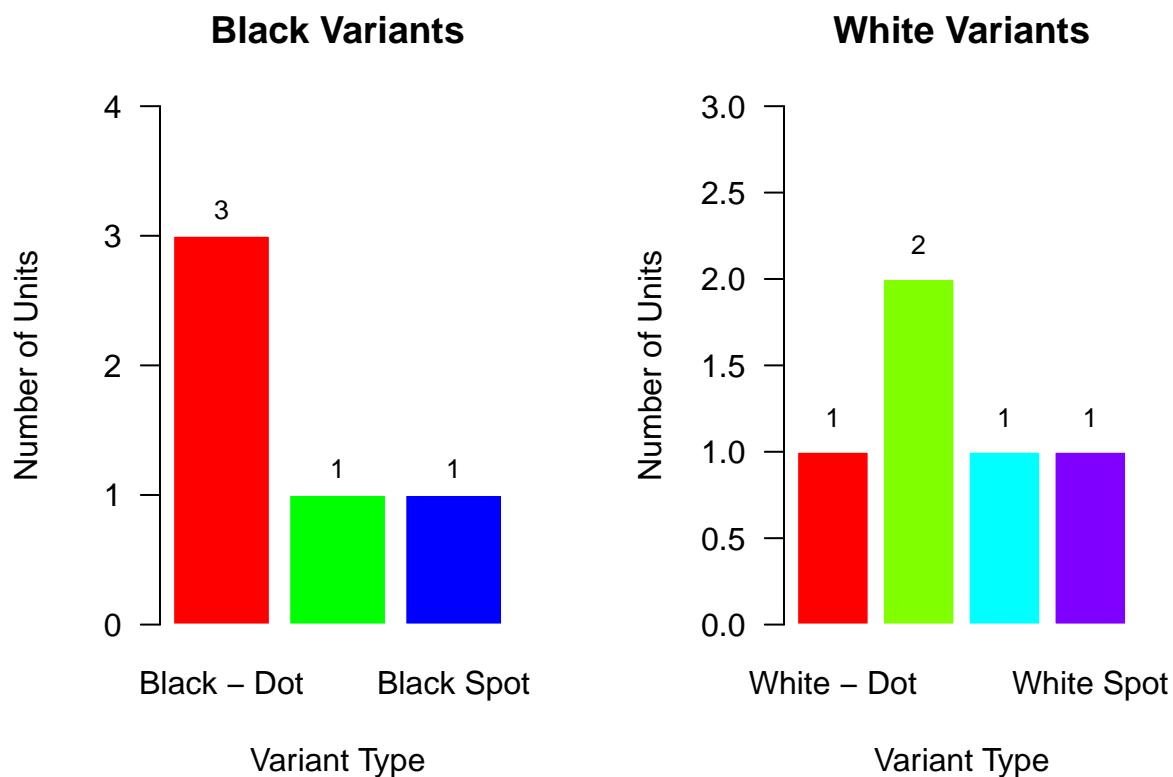
text(x = bar_black, y = black$n + 0.2, labels = black$n, cex = 0.8, col = "black")
#white
bar_white <- barplot(white$n,
                      names.arg = white$variation,
                      main = "White Variants",
                      xlab = "Variant Type",
                      ylab = "Number of 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))
```