

RWorksheet_Nava#4b

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2024-10-28

Using Loop Function 1.

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

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

for (i in 1:5) {
  for (j in 1:5) {

    matrixResult[i, j] <- abs(vectorA[i] - vectorA[j])
  }
}

print(matrixResult)
```

```
##      [,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.

```
numRows <- 5

for (i in 1:numRows) {
  for (j in 1:i) {
    cat('*', " ")
  }
  cat("\n")
}
```

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

3.

```
startNum <- 0
```

```

if (is.na(startNum) || startNum < 0) {
  stop("Please enter a valid non-negative number.")
}

fib1 <- 0
fib2 <- 1

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

## Fibonacci sequence starting from 0 up to 500:
repeat {
  if (fib1 >= startNum) {
    cat(fib1, "\n")
  }
  nextFib <- fib1 + fib2
  fib1 <- fib2
  fib2 <- nextFib

  if (fib1 > 500) {
    break
  }
}

```

```

## 0
## 1
## 1
## 2
## 3
## 5
## 8
## 13
## 21
## 34
## 55
## 89
## 144
## 233
## 377

```

Using Basic Graphics (plot(), barplot(), pie(), hist()) 4. a.

```

library(openxlsx)

dataset <- read.xlsx("household_data2.xlsx", sheet = 1)

print(head(dataset))

```

```

##   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
```

```
print(str(dataset))
```

```
## 'data.frame':   28 obs. of  3 variables:
## $ Shoe.Size: num  6.5 9 8.5 8.5 10.5 7 9.5 9 13 7.5 ...
## $ Height   : num  66 68 64.5 65 70 64 70 71 72 64 ...
## $ Gender    : chr  "F" "F" "F" "F" ...
## NULL
```

```
dataset$Gender <- as.character(dataset$Gender) # Convert to character if needed
```

b.

```
library(openxlsx)
```

```
dataset <- read.xlsx("household_data2.xlsx", sheet = 1)
```

```
print(head(dataset))
```

```
##   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
```

```
uniqueValues <- unique(dataset$Gender)
print(uniqueValues)
```

```
## [1] "F" "M"
```

```
male_subset <- subset(dataset, Gender == "M")
female_subset <- subset(dataset, Gender == "F")
```

```
numMales <- nrow(male_subset)
numFemales <- nrow(female_subset)
```

```
cat("Number of observations in Male:", numMales, "\n")
```

```
## Number of observations in Male: 14
```

```
cat("Number of observations in Female:", numFemales, "\n")
```

```
## Number of observations in Female: 14
```

5.

```
categories <- c("Food", "Electricity", "Savings", "Miscellaneous")
expenses <- c(60, 10, 5, 25)
```

```
percentages <- round((expenses / sum(expenses)) * 100)
```

```

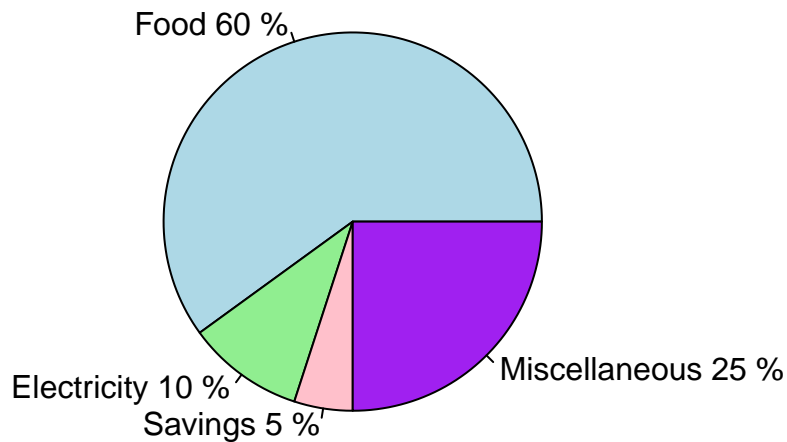
labels <- paste(categories, percentages, "%", sep = " ")

colors <- c("lightblue", "lightgreen", "pink", "purple")

# Create the pie chart
pie(expenses, labels = labels, col = colors, main = "Monthly Income Expenses of Dela Cruz Family")

```

Monthly Income Expenses of Dela Cruz Family



6. 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 0.2 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 ...

```

b.

```

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

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

```

c.

```

species_count <- table(iris$Species)

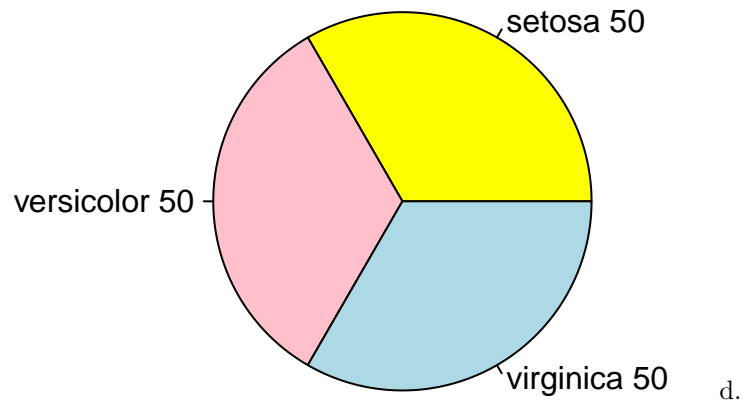
labels <- paste(names(species_count), species_count)

colors <- c("yellow", "pink", "lightblue")

```

```
pie(species_count, labels = labels, col = colors, main = "Species Distribution in Iris Dataset")
```

Species Distribution in Iris Dataset



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

```
lastSixSetosa <- tail(setosa, 6)
lastSixVersicolor <- tail(versicolor, 6)
lastSixVirginica <- tail(virginica, 6)
```

lastSixSetosa

##	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

lastSixVersicolor

##	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

lastSixVirginica

##	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

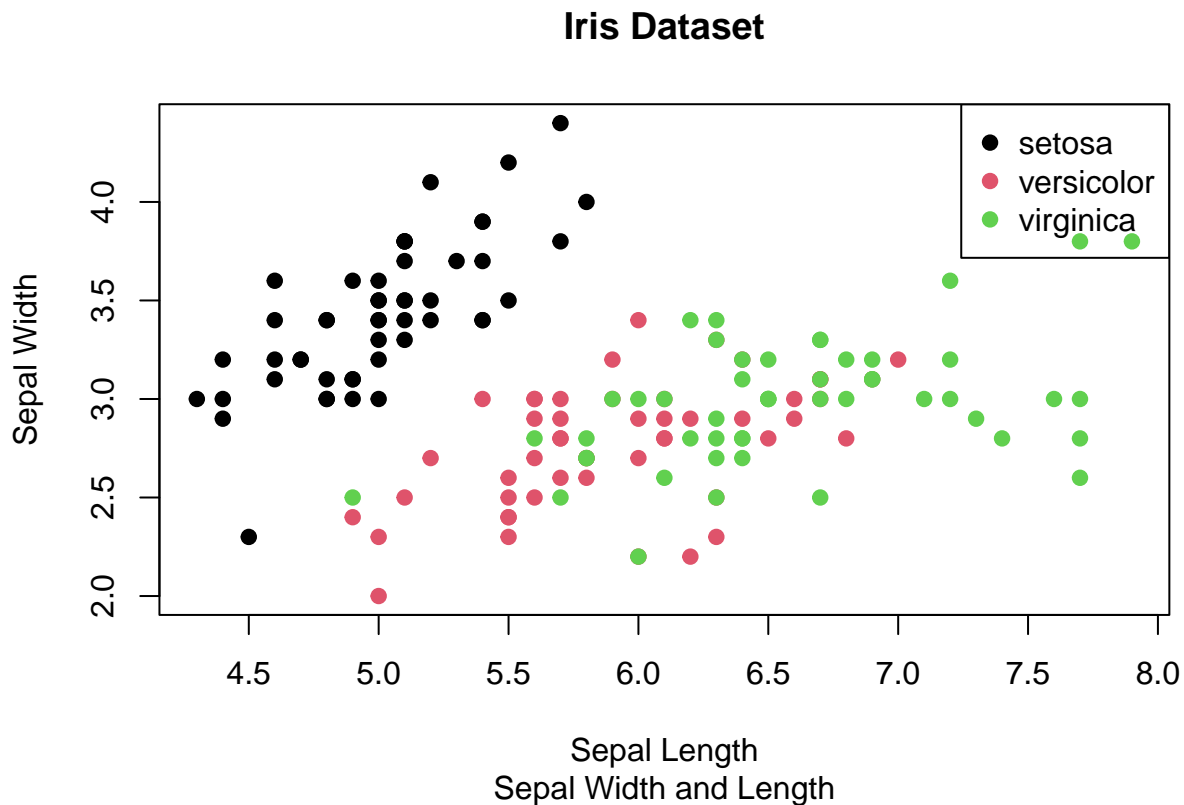
```
## 148      6.5      3.0      5.2      2.0 virginica
## 149      6.2      3.4      5.4      2.3 virginica
## 150      5.9      3.0      5.1      1.8 virginica
```

e.

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

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

legend("topright", legend = levels(iris$Species), col = 1:3, pch = 19)
```



f.

The scatter plot is a display of the variation within the iris species for changes in sepal length and width. You can see there is clustering - where some species are very close to another, hence similar in measurement. For example, setosa may typically be distinguished from the rest, say versicolor and virginica, with smaller sepal length and width. This might also suggest that setosa may be somewhat different from the other two species, whose measures overlap. The means computed in b give a summary measure for each measurement, providing the mean central tendency of each characteristic for all of the species. In this case, the higher mean values in the versicolor and virginica species for the characteristics Petal.Length and Petal.Width indicate that these species have larger petals compared to setosa.

7.

```
library(openxlsx)
```

```

alex_data <- read.xlsx("alex_data.xlsx")

print("Original Dataset:")

## [1] "Original Dataset:"
print(head(alex_data)) # Show the first 6 rows

##      Variants Count
## 1 Black Dot      10
## 2 Black Plus     15
## 3 Black Show      5
## 4 Black Spot      8
## 5 White Dot      12
## 6 White Plus     20

alex_data$Variants <- gsub("^\\s+|\\s+$", "", alex_data$Variants) # Trim leading and trailing spaces
alex_data$Variants <- gsub("\\s+", " ", alex_data$Variants) # Replace multiple spaces with a single space

alex_data$Variants <- gsub("Black Dot", "BlackDot", alex_data$Variants)
alex_data$Variants <- gsub("Black Plus", "BlackPlus", alex_data$Variants)
alex_data$Variants <- gsub("Black Show", "BlackShow", alex_data$Variants)
alex_data$Variants <- gsub("Black Spot", "BlackSpot", alex_data$Variants)
alex_data$Variants <- gsub("White Dot", "WhiteDot", alex_data$Variants)
alex_data$Variants <- gsub("White Plus", "WhitePlus", alex_data$Variants)
alex_data$Variants <- gsub("White Show", "WhiteShow", alex_data$Variants)
alex_data$Variants <- gsub("White Spot", "WhiteSpot", alex_data$Variants)

print("Cleaned Dataset Snippet:")

## [1] "Cleaned Dataset Snippet:"
print(head(alex_data))

##      Variants Count
## 1 BlackDot      10
## 2 BlackPlus     15
## 3 BlackShow      5
## 4 BlackSpot      8
## 5 WhiteDot      12
## 6 WhitePlus     20

write.xlsx(alex_data, "cleaned_alex_data.xlsx")

```

b.

```

library(dplyr)

variationCounts <- alex_data %>%
  count(Variants)

save(variationCounts, file = "variations.RData")

```

```
print("Variation Counts:")
```

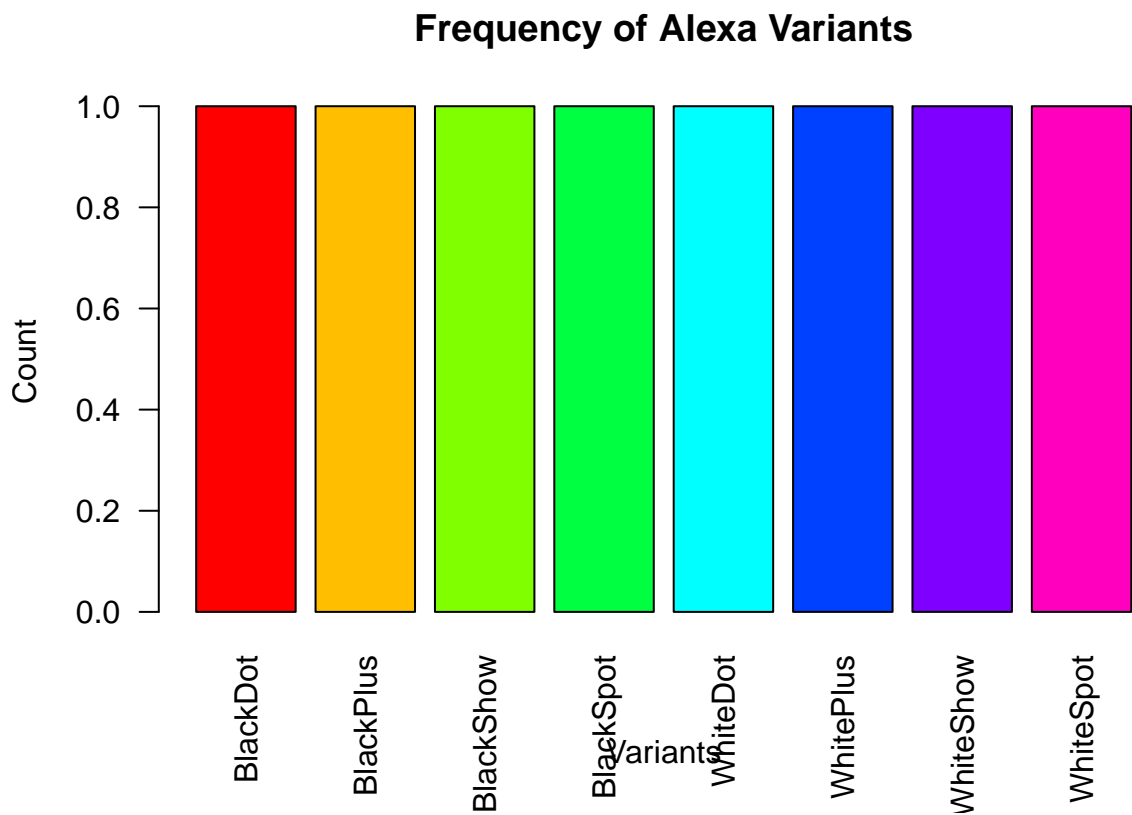
```
## [1] "Variation Counts:"
```

```
print(variationCounts)
```

```
##      Variants n
## 1 BlackDot 1
## 2 BlackPlus 1
## 3 BlackShow 1
## 4 BlackSpot 1
## 5 WhiteDot 1
## 6 WhitePlus 1
## 7 WhiteShow 1
## 8 WhiteSpot 1
```

c.

```
barplot(variationCounts$n,
        names.arg = variationCounts$Variants,
        col = rainbow(nrow(variationCounts)),
        main = "Frequency of Alexa Variants",
        xlab = "Variants",
        ylab = "Count",
        las = 2)
```

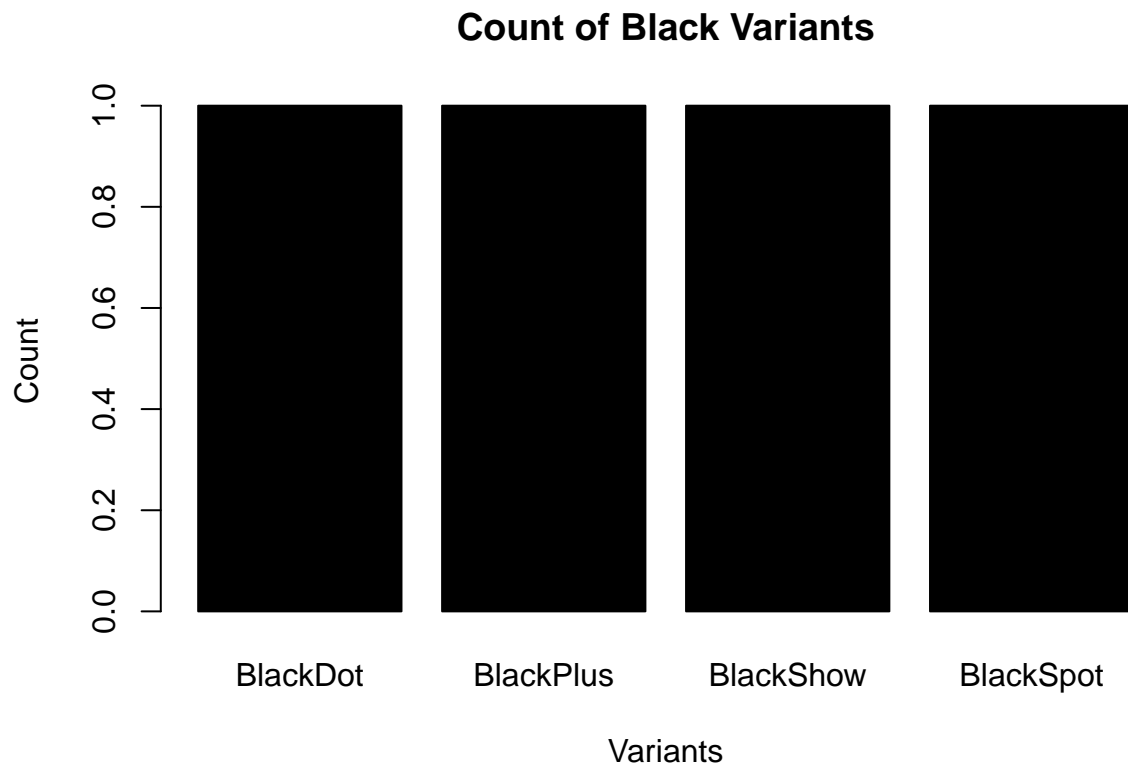


d.

```
# Step 5: Separate counts for black and white variations
blackVariations <- variationCounts[grep("Black", variationCounts$Variants), ]
whiteVariations <- variationCounts[grep("White", variationCounts$Variants), ]
```



```
# Barplot for Black Variants
barplot(blackVariations$n,
        names.arg = blackVariations$Variants,
        col = "black",
        main = "Count of Black Variants",
        xlab = "Variants",
        ylab = "Count")
```



```
# Barplot for White Variants
barplot(whiteVariations$n,
        names.arg = whiteVariations$Variants,
        col = "lightgrey",
        main = "Count of White Variants",
        xlab = "Variants",
        ylab = "Count")
```

