

Practical 2

EXERCISE 1

```
# Load the dataset
titanic <- read.csv("./datasets/titanic.csv", header = TRUE, sep = ",")
```

```
# Remove the row index column
titanic <- subset(titanic, select = -X)
```

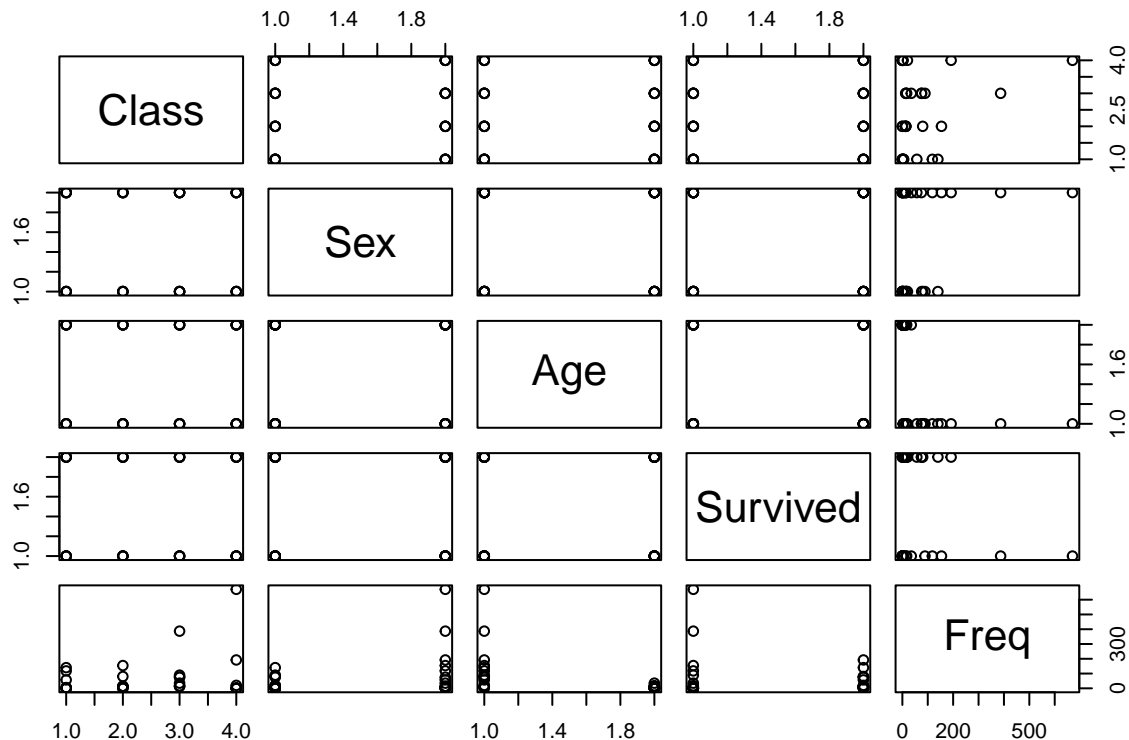
```
# Display the first few rows of the dataset
head(titanic)
```

```
##   Class   Sex   Age Survived Freq
## 1   1st  Male Child      No    0
## 2   2nd  Male Child      No    0
## 3   3rd  Male Child      No   35
## 4  Crew  Male Child      No    0
## 5   1st Female Child      No    0
## 6   2nd Female Child      No    0
```

```
# Display the summary statistics of the dataset
summary(titanic)
```

```
##      Class           Sex           Age           Survived
## Length:32      Length:32      Length:32      Length:32
## Class :character  Class :character  Class :character  Class :character
## Mode  :character  Mode  :character  Mode  :character  Mode  :character
##
##
##
##      Freq
## Min.   : 0.00
## 1st Qu.: 0.75
## Median :13.50
## Mean   :68.78
## 3rd Qu.:77.00
## Max.   :670.00
```

```
# Create a scatterplot matrix of the dataset
plot(titanic)
```



```
# Using str() function we can see the structure of the dataset
str(titanic)
```

```
## 'data.frame':  32 obs. of  5 variables:
## $ Class   : chr  "1st" "2nd" "3rd" "Crew" ...
## $ Sex     : chr  "Male" "Male" "Male" "Male" ...
## $ Age     : chr  "Child" "Child" "Child" "Child" ...
## $ Survived: chr  "No" "No" "No" "No" ...
## $ Freq    : int   0 0 35 0 0 0 17 0 118 154 ...
```

Quantitative (Numerical) Variables:

- Freq: Represents the frequency (number of people)

Categorical Variables:

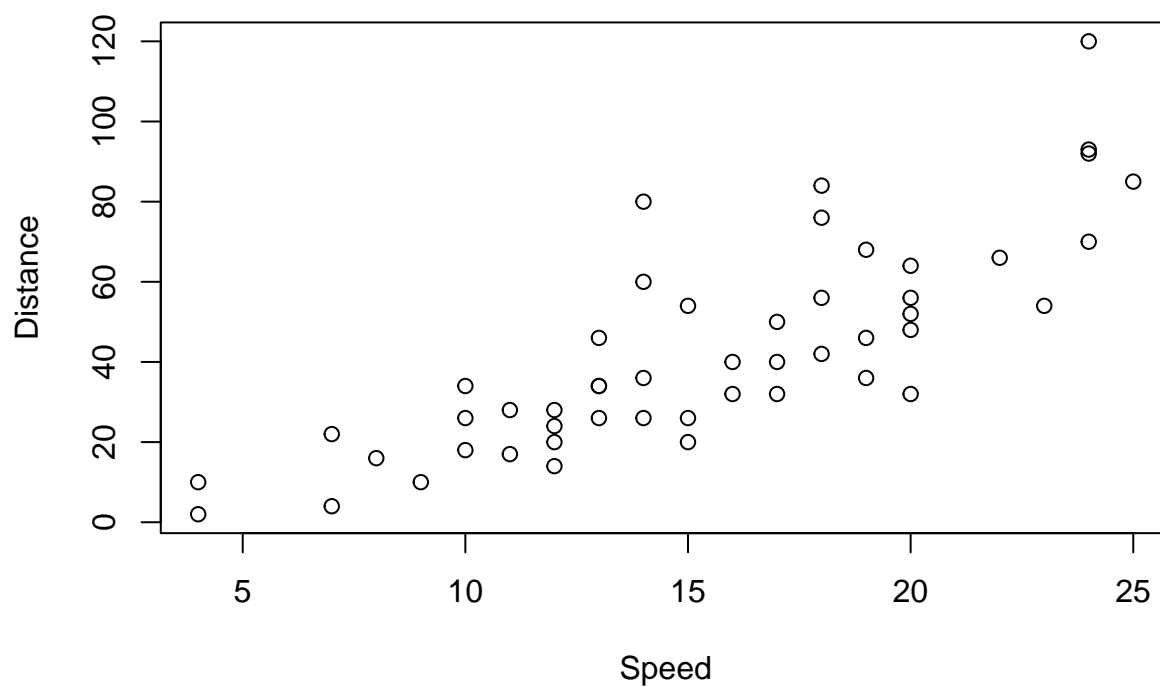
- X: Row index (treated as categorical, despite being an integer)
- Class: Passenger class (e.g., "1st", "2nd", "3rd", "Crew")
- Sex: Gender (e.g., "Male", "Female")
- Age: Age group (e.g., "Child", "Adult")
- Survived: Survival status (e.g., "Yes", "No")

EXERCISE 2

```
# Load the dataset
cars <- read.csv("./datasets/cars.csv", header = TRUE, sep = ",")

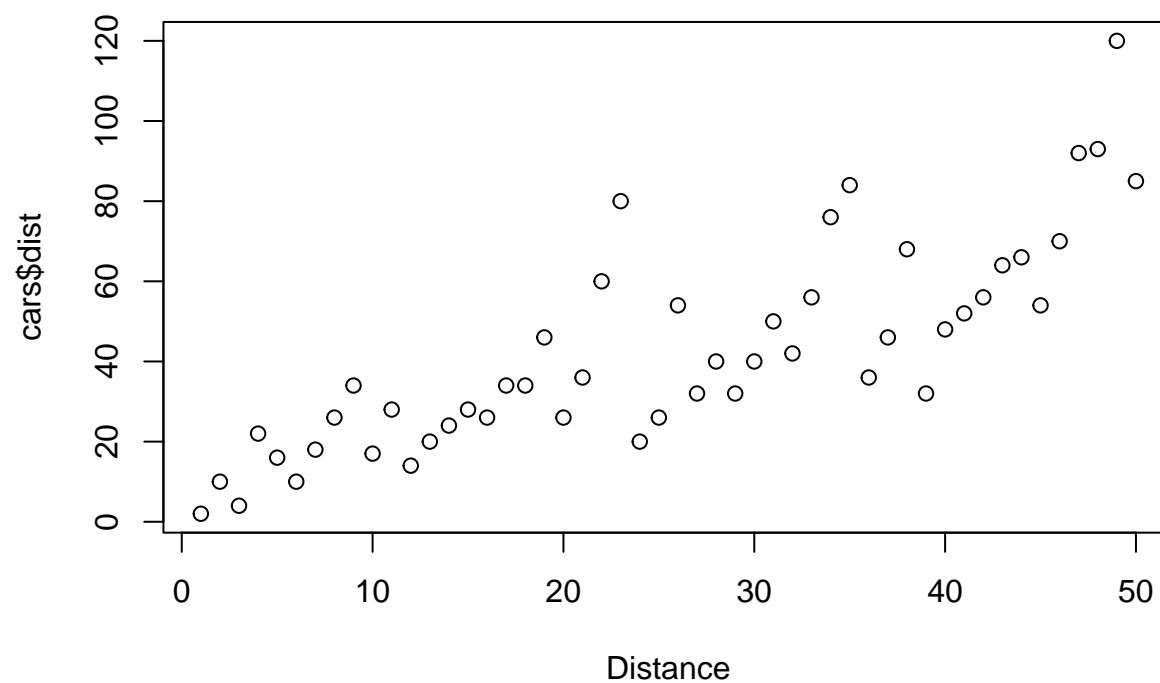
# Make a plot of the distance field in terms of the speed field
plot(cars$speed, cars$dist, xlab = "Speed", ylab = "Distance",
     main = "Distance vs. Speed")
```

Distance vs. Speed



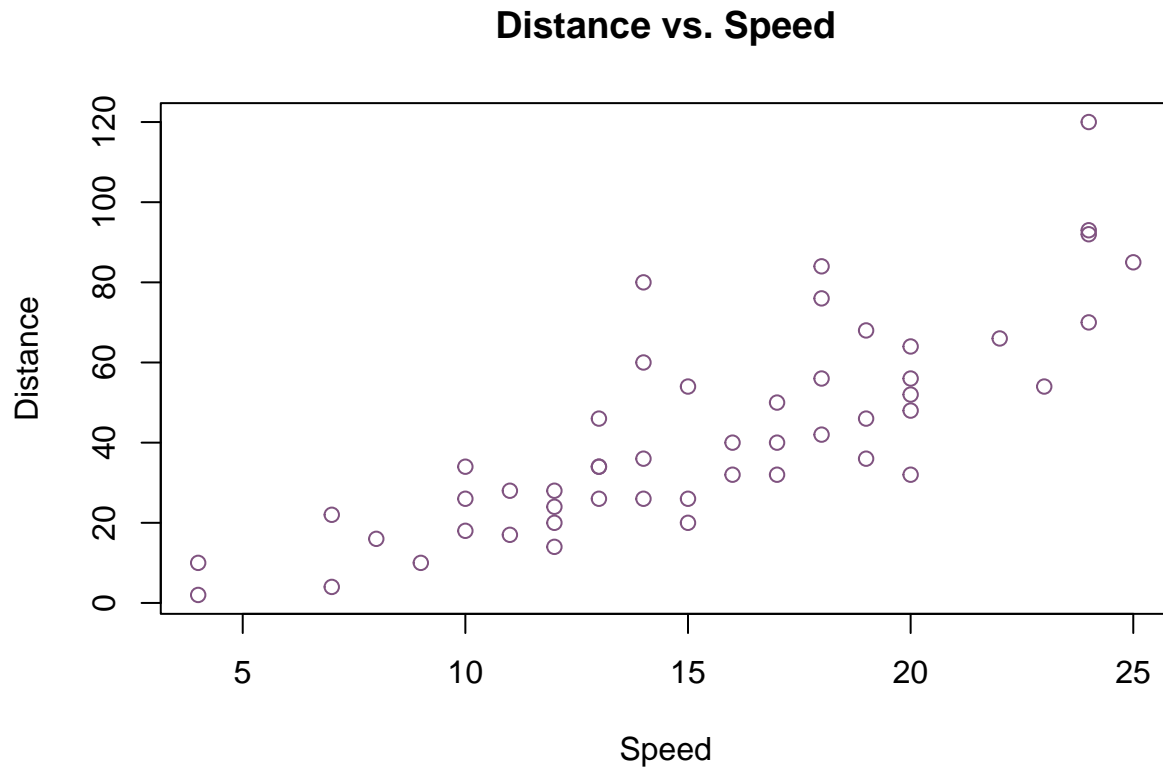
```
# Create a histogram of the distance field  
plot(cars$dist, xlab = "Distance", main = "Histogram of Distance")
```

Histogram of Distance



```
pdf("../documents/exercice2/distance_vs_speed_plot.pdf")
```

```
# Create a scatterplot of the speed field
plot(cars$speed, cars$dist,
     xlab = "Speed",
     ylab = "Distance",
     main = "Distance vs. Speed",
     col = "#805380")
```



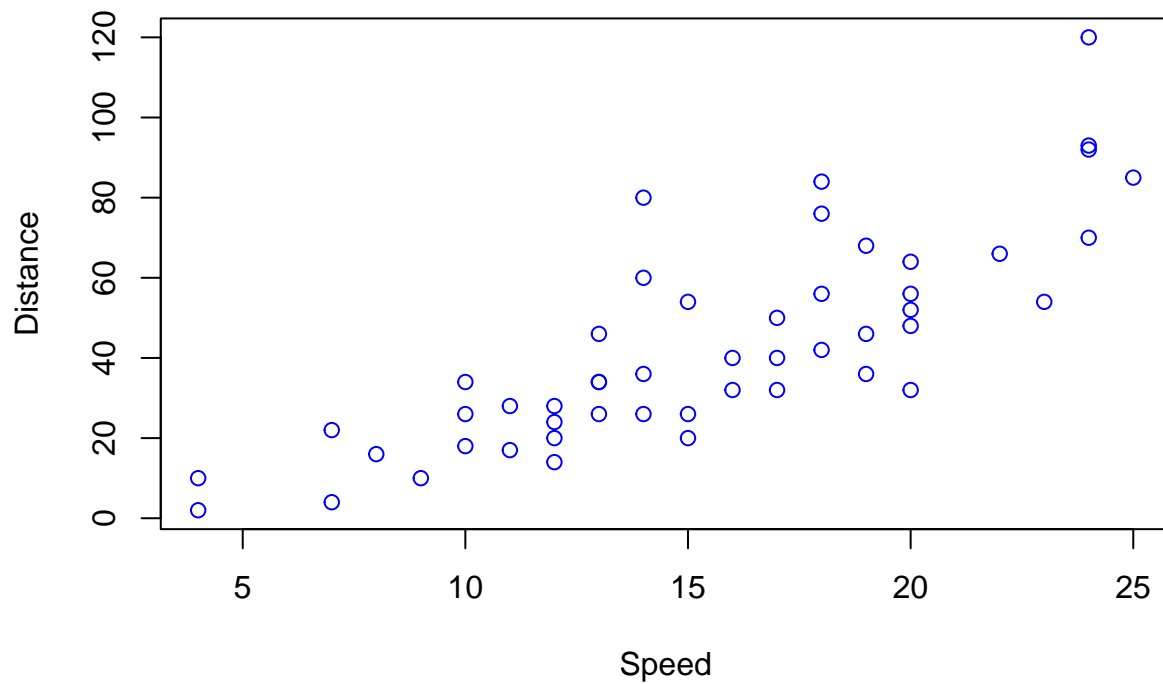
```
dev.off()
```

```
## pdf
## 3
```

```
# Save the scatterplot to a PDF file
pdf("./documents/exercice2/modified_distance_vs_speed_plot.pdf")
```

```
# Scatterplot of Distance vs Speed with modified title, axis labels, and color
plot(cars$speed, cars$dist,
     xlab = "Speed",
     ylab = "Distance",
     main = "Distance vs Speed",
     col = "blue")
```

Distance vs Speed



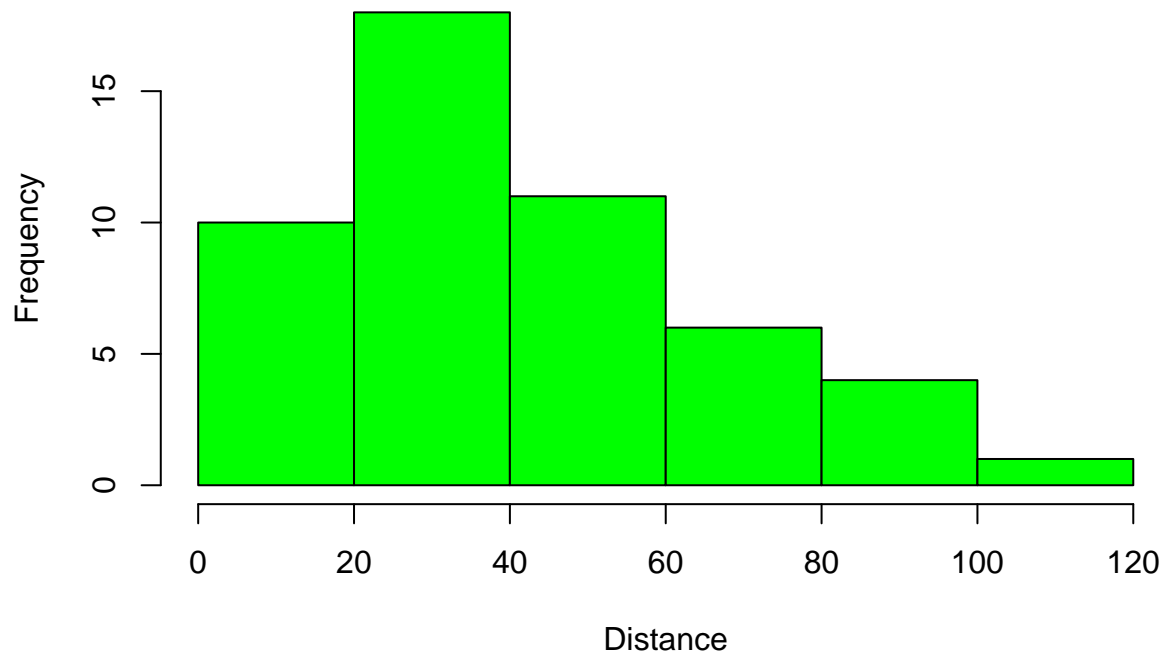
```
dev.off()
```

```
## pdf  
## 3
```

```
# Save the histogram of the distance field to a PDF file  
pdf("../documents/exercice2/ex2_histogram_distance.pdf")
```

```
# Histogram of Distance with modified title, axis labels, and color  
hist(cars$dist,  
      xlab = "Distance",  
      main = "Histogram of Distance",  
      col = "green",  
      col.main = "red")
```

Histogram of Distance



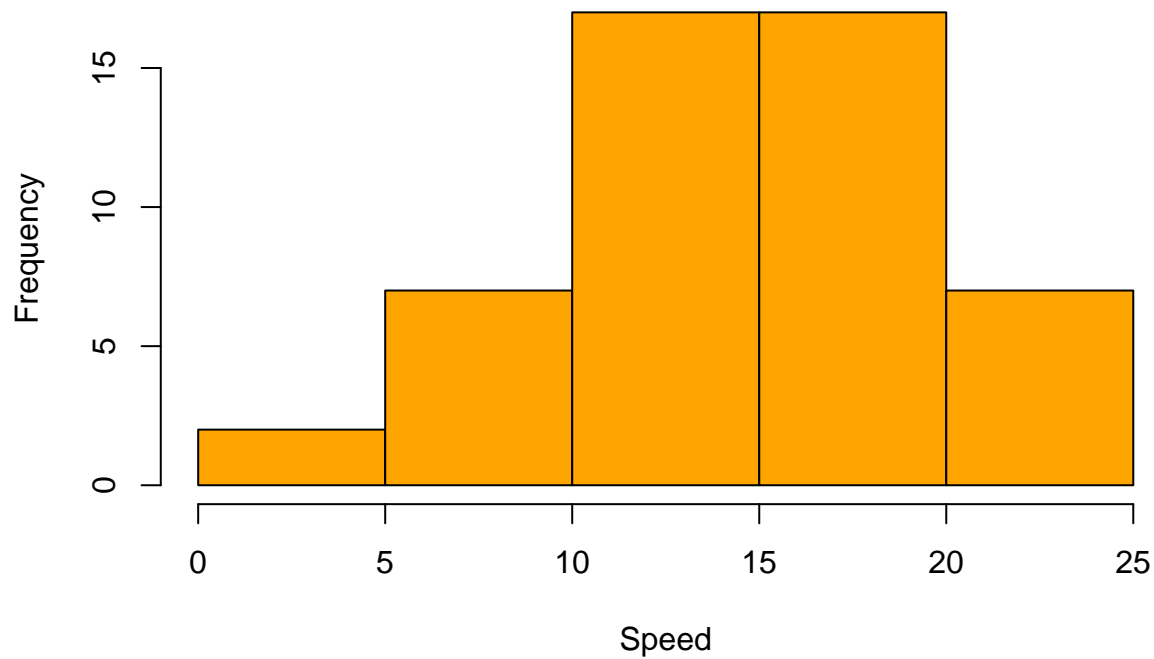
```
dev.off()
```

```
## pdf  
## 3
```

```
# Save the histogram of the speed field to a PDF file  
pdf("../documents/exercice2/ex2_histogram_speed.pdf")
```

```
# Histogram of Speed with modified title, axis labels, and color  
hist(cars$speed,  
     xlab = "Speed",  
     main = "Histogram of Speed",  
     col = "orange",  
     col.main = "blue")
```

Histogram of Speed



```
dev.off()
```

```
## pdf  
## 3
```

EXERCISE 3

```
# Load the dataset  
cars <- read.csv("./datasets/cars.csv", header = TRUE, sep = ",")
```

```
# Remove the first column of the cars data frame  
cars <- cars[, -1]
```

```
# Construct a new data frame  
new_cars <- data.frame(speed = c(21, 34), dist = c(47, 87))
```

```
# Add the constructed data frame to the cars data frame  
cars <- rbind(cars, new_cars)
```

```
# Sort the data in the resulting dataset by column speed (ascending)  
cars <- cars[order(cars$speed), ]
```

```
# Write the resulting dataset to a CSV file  
write.csv(cars, file = "./datasets/cars_sorted.csv", row.names = FALSE)
```

EXERCISE 4

```
# Load the dataset  
airquality <- read.csv("./datasets/airquality.csv", header = TRUE, sep = ",")
```

```

# Display the first two rows of the dataset
print(airquality[1:2, ])

##   Ozone Solar.R Wind Temp Month Day
## 1    41     190  7.4   67     5   1
## 2    36     118  8.0   72     5   2

# How many rows are in the dataset?
nrow(airquality)

## [1] 153

# What is the value of Ozone in the 40th row?
airquality[40, "Ozone"]

## [1] 71

# How many missing values are there in the Ozone column?
sum(is.na(airquality$Ozone))

## [1] 37

# What is the mean of the Ozone column in this dataset? Exclude NA values
airquality <- read.csv("./datasets/airquality.csv", header = TRUE, sep = ",")
ozone_clean <- na.omit(airquality$Ozone)
print(mean(ozone_clean))

## [1] 42.12931

# Extract the rows where the Ozone value is greater than 31
# and Temp value is greater than 90
airquality <- read.csv("./datasets/airquality.csv", header = TRUE, sep = ",")
airquality <- na.omit(airquality)
airquality_subset <- airquality[airquality$Ozone > 31 & airquality$Temp > 90, ]

# What is the mean of Solar.R in this subset?
print(mean(airquality_subset$Solar.R))

## [1] 212.8

```

EXERCISE 5

```

# Discretize the Ozone column into 5 bins
aux <- cut(airquality$Ozone,
           breaks = 5,
           labels = c("bin1", "bin2", "bin3", "bin4", "bin5"))
# Add NA to the levels
aux <- addNA(aux)
print(aux)

##   [1] bin2 bin2 bin1 bin1 bin1 bin1 bin1 bin1 bin1 bin1 bin1 bin1 bin1 bin1 bin1
##  [16] bin1 bin1 bin1 bin1 bin1 bin1 bin2 bin4 bin2 bin1 bin3 bin2 bin1 bin1 bin2
##  [31] bin1 bin1 bin1 bin5 bin2 bin1 bin2 bin2 bin3 bin3 bin3 bin3 bin1 bin1 bin1
##  [46] bin2 bin2 bin2 bin3 bin2 bin1 bin3 bin4 bin1 bin2 bin3 bin2 bin2 bin2 bin2
##  [61] bin1 bin1 bin4 bin3 bin4 bin2 bin1 bin2 bin1 bin2 bin1 bin1 bin2 bin1 bin1
##  [76] bin2 bin5 bin3 bin3 bin4 bin3 bin3 bin3 bin3 bin3 bin3 bin2 bin1 bin1 bin1
##  [91] bin1 bin1 bin2 bin1 bin1 bin1 bin1 bin2 bin1 bin1 bin1 bin1 bin1 bin1 bin2

```



```
## [106] bin1 bin1 bin1 bin1 bin1 bin1
## Levels: bin1 bin2 bin3 bin4 bin5 <NA>

# Discretize the Solar.R column into 4 bins
aux <- cut(airquality$Solar.R,
           breaks = 4,
           labels = c("bin1", "bin2", "bin3", "bin4"))

# Add NA to the levels
aux <- addNA(aux)
print(aux)

## [1] bin3 bin2 bin2 bin4 bin4 bin2 bin1 bin4 bin4 bin4 bin1 bin4 bin4 bin1 bin4
## [16] bin1 bin1 bin4 bin1 bin2 bin1 bin3 bin3 bin4 bin2 bin4 bin4 bin2 bin3 bin4
## [31] bin1 bin2 bin2 bin4 bin3 bin3 bin3 bin4 bin4 bin4 bin4 bin3 bin4 bin3 bin1
## [46] bin4 bin4 bin4 bin3 bin3 bin1 bin4 bin3 bin1 bin1 bin3 bin4 bin4 bin4 bin1
## [61] bin1 bin1 bin4 bin3 bin3 bin3 bin4 bin2 bin1 bin1 bin2 bin3 bin3 bin4 bin1
## [76] bin3 bin3 bin3 bin3 bin3 bin3 bin3 bin2 bin3 bin3 bin3 bin2 bin2 bin3 bin3
## [91] bin3 bin4 bin3 bin4 bin3 bin1 bin2 bin3 bin3 bin1 bin3 bin3 bin3 bin1 bin2
## [106] bin1 bin1 bin3 bin3 bin2 bin3
## Levels: bin1 bin2 bin3 bin4 <NA>

# Load the dataset
airquality <- read.csv("./datasets/airquality.csv", header = TRUE, sep = ",")

# Create a new column called "cumulative_days"
cumulative_days <- c(0, 31, 61, 92, 123)

# Add the new column to the dataset and adjust the index (May is month 5)
airquality$AbsDay <- airquality$Day + cumulative_days[airquality$Month - 4]

# Display the updated dataset
head(airquality)

## Ozone Solar.R Wind Temp Month Day AbsDay
## 1 41 190 7.4 67 5 1 1
## 2 36 118 8.0 72 5 2 2
## 3 12 149 12.6 74 5 3 3
## 4 18 313 11.5 62 5 4 4
## 5 NA NA 14.3 56 5 5 5
## 6 28 NA 14.9 66 5 6 6
```

EXERCISE 6

```
# Load the dataset
titanic <- read.csv("./datasets/titanic.csv", header = TRUE, sep = ",")
titanic$Class <- as.numeric(factor(titanic$Class))

# Numerize the Class column
print(titanic$Class <- as.numeric(titanic$Class))

## [1] 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4

# Load the dataset
titanic <- read.csv("./datasets/titanic.csv", header = TRUE, sep = ",")
```

```
# Create a new data frame (titanic2) by expanding rows based on the Freq column
titanic2 <- titanic[rep(seq_len(nrow(titanic)), titanic$Freq), ]
head(titanic2)
```

```
##      X Class Sex   Age Survived Freq
## 3     3   3rd Male Child      No   35
## 3.1 3     3rd Male Child      No   35
## 3.2 3     3rd Male Child      No   35
## 3.3 3     3rd Male Child      No   35
## 3.4 3     3rd Male Child      No   35
## 3.5 3     3rd Male Child      No   35
```

```
# Load the dataset
```

```
titanic <- read.csv("./datasets/titanic.csv", header = TRUE, sep = ",")
```

```
# Define the colors for the bar plots
```

```
colors <- c("orange", "#9370DB", "blue", "darkgrey")
```

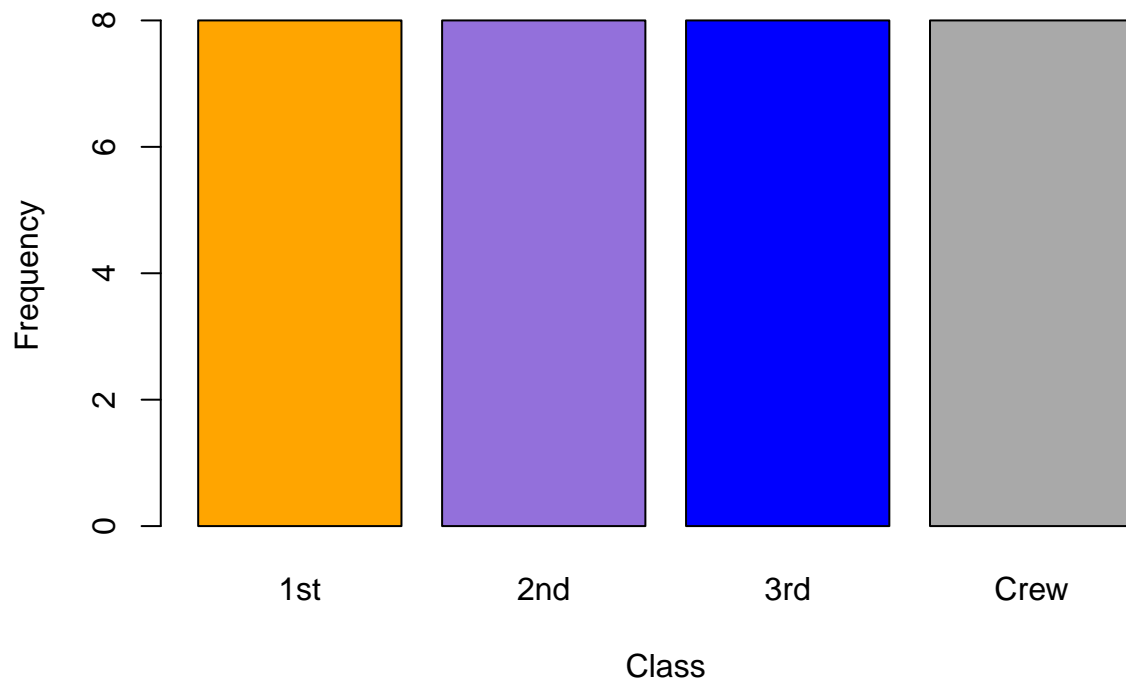
```
# Plot distribution of Class in the original dataset with multiple colors
```

```
barplot(table(titanic$Class),
        main = "Class Distribution in Original Titanic Data",
        xlab = "Class",
        ylab = "Frequency",
        col = colors)
```

```
# Plot distribution of Class in the original dataset
```

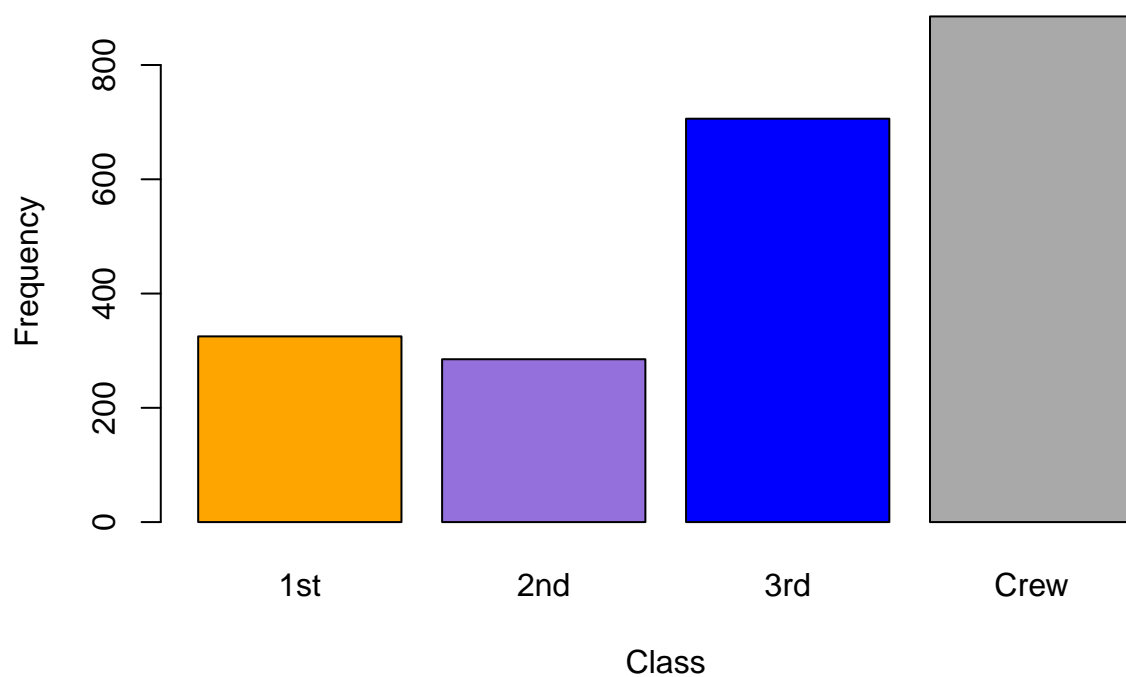
```
barplot(table(titanic$Class),
        main = "Class Distribution in Original Titanic Data",
        xlab = "Class",
        ylab = "Frequency",
        col = colors)
```

Class Distribution in Original Titanic Data



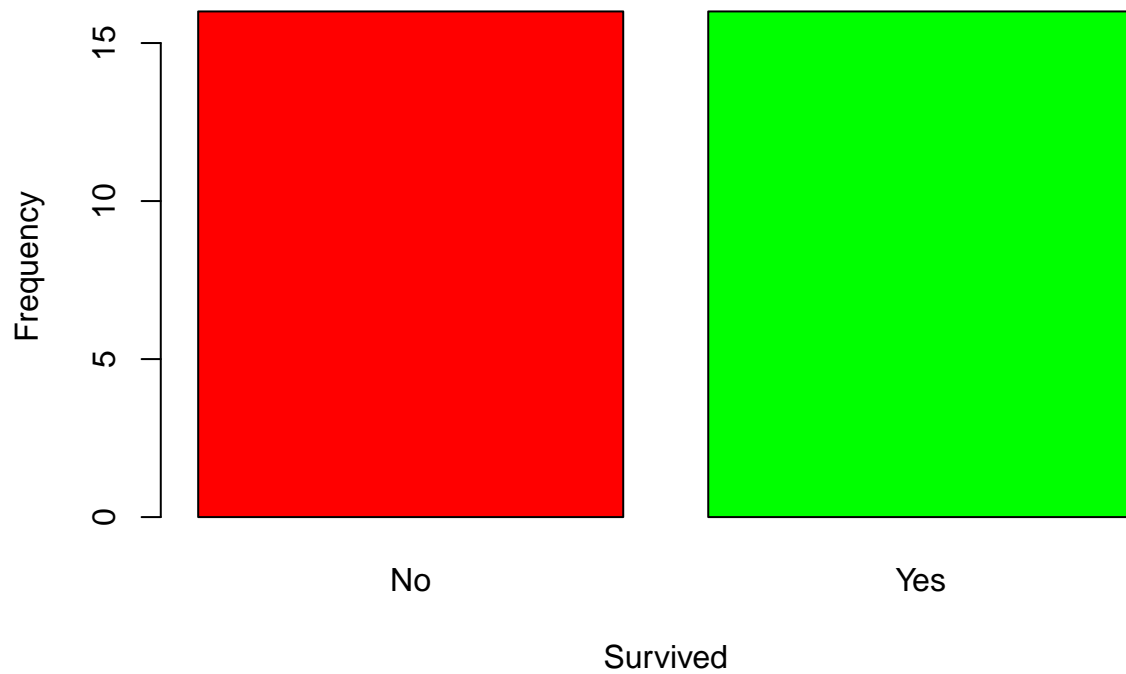
```
# Plot distribution of Class in the new dataset (titanic2)
barplot(table(titanic2$Class),
        main = "Class Distribution in Expanded Titanic Data",
        xlab = "Class",
        ylab = "Frequency",
        col = colors)
```

Class Distribution in Expanded Titanic Data



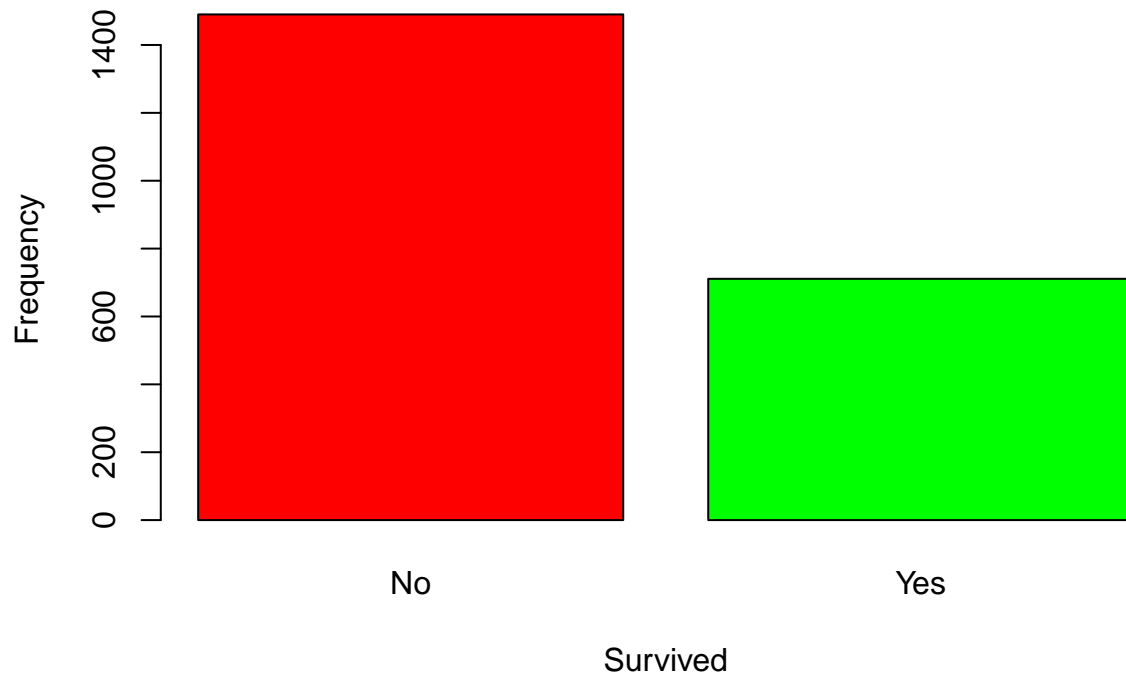
```
# Plot distribution of Survival in the original dataset
barplot(table(titanic$Survived),
  main = "Survival Distribution in Original Titanic Data",
  xlab = "Survived",
  ylab = "Frequency",
  col = c("red", "green"))
```

Survival Distribution in Original Titanic Data



```
# Plot distribution of Survival in the new dataset (titanic2)
barplot(table(titanic2$Survived),
  main = "Survival Distribution in Expanded Titanic Data",
  xlab = "Survived",
  ylab = "Frequency",
  col = c("red", "green"))
```

Survival Distribution in Expanded Titanic Data



```
# Correlation between Class and Survival in the original dataset
# Create a contingency table
class_survived_table <- table(titanic$Class, titanic$Survived)

# Calculate survival rates by class
survival_rates <- prop.table(class_survived_table, margin = 1)

# Print the contingency table and survival rates
print(class_survived_table)
```

```
##
##      No Yes
## 1st   4   4
## 2nd   4   4
## 3rd   4   4
## Crew  4   4
```

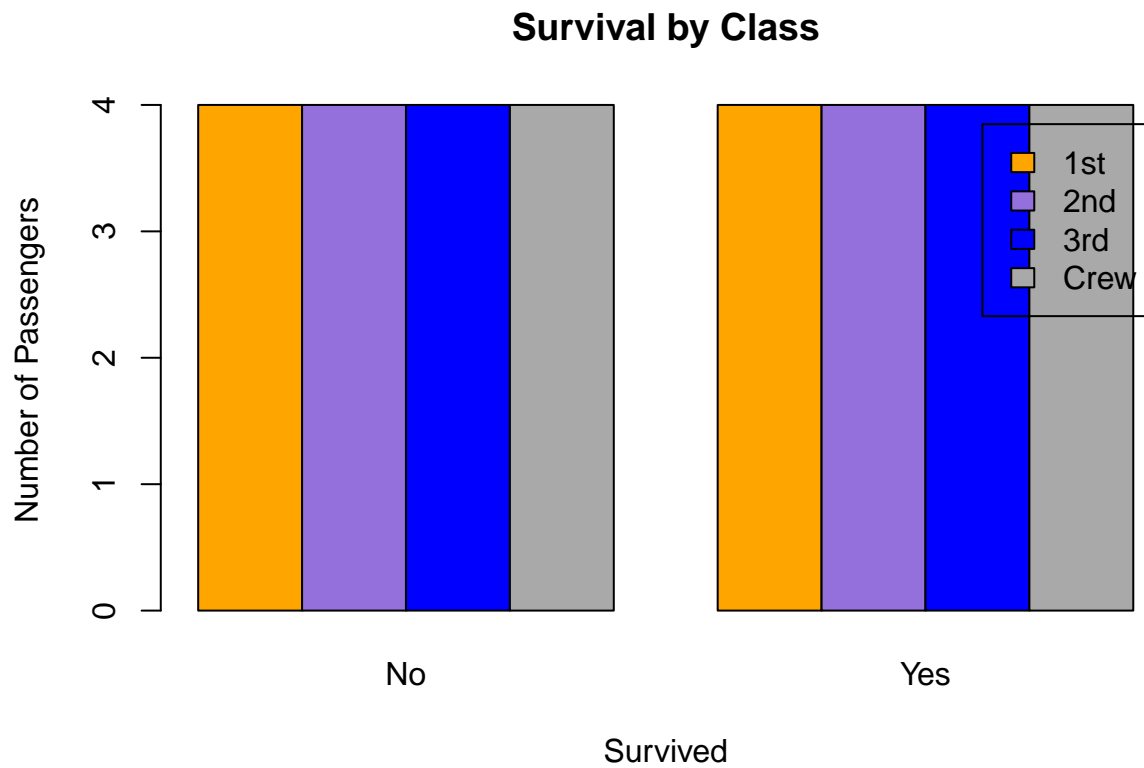
```
print(survival_rates)
```

```
##
##      No Yes
## 1st 0.5 0.5
## 2nd 0.5 0.5
## 3rd 0.5 0.5
## Crew 0.5 0.5
```

```
# Visualizing the relationship

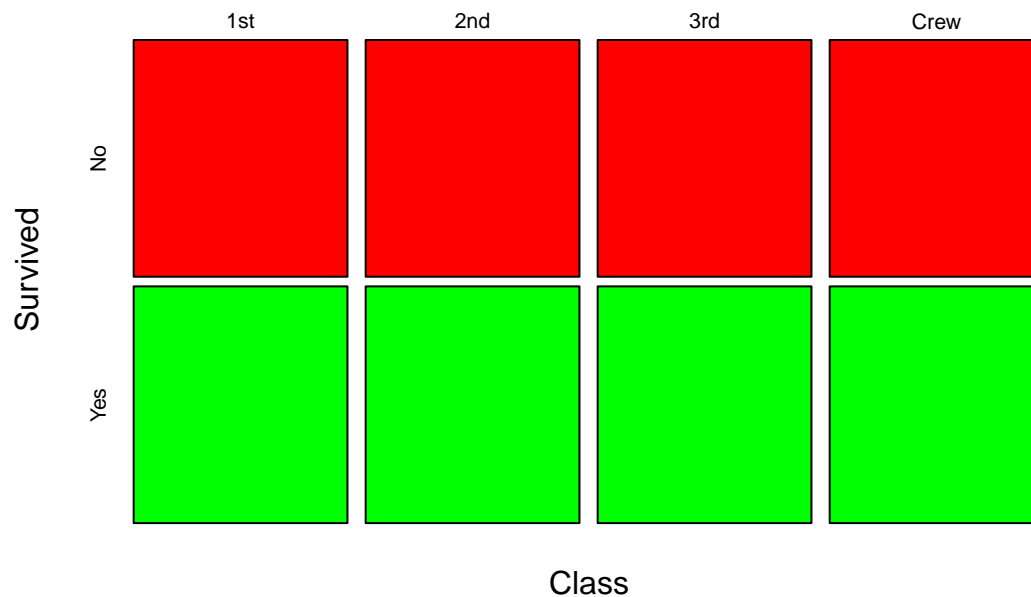
# Bar plot of survival by class
barplot(class_survived_table,
        beside = TRUE,
```

```
col = colors,
legend = rownames(class_survived_table),
main = "Survival by Class",
xlab = "Survived",
ylab = "Number of Passengers")
```



```
# Mosaic plot of Class vs Survived
mosaicplot(~ Class + Survived,
  data = titanic,
  col = c("red", "green"),
  main = "Mosaic Plot of Class vs Survived")
```

Mosaic Plot of Class vs Survived



```
# Correlation between Class and Survival in the new dataset (titanic2)
```

```
# Create a contingency table for titanic2
```

```
class_survived_table2 <- table(titanic2$Class, titanic2$Survived)
```

```
# Calculate survival rates by class for titanic2
```

```
survival_rates2 <- prop.table(class_survived_table2, margin = 1)
```

```
# Print the contingency table and survival rates for titanic2
```

```
print(class_survived_table2)
```

```
##
```

```
##      No Yes
```

```
## 1st 122 203
```

```
## 2nd 167 118
```

```
## 3rd 528 178
```

```
## Crew 673 212
```

```
print(survival_rates2)
```

```
##
```

```
##      No      Yes
```

```
## 1st 0.3753846 0.6246154
```

```
## 2nd 0.5859649 0.4140351
```

```
## 3rd 0.7478754 0.2521246
```

```
## Crew 0.7604520 0.2395480
```

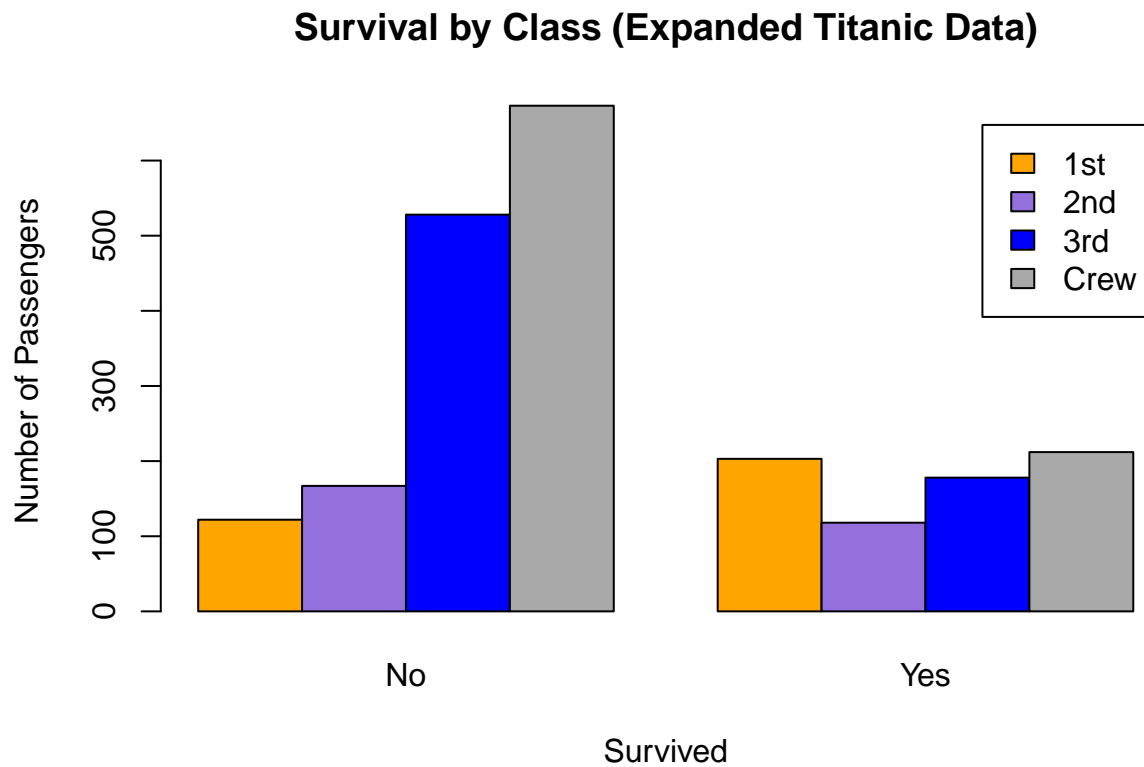
```
# Visualizing the relationship in titanic2
```

```
# Bar plot of survival by class in titanic2
```

```
barplot(class_survived_table2,
        beside = TRUE,
        col = colors,
        legend = rownames(class_survived_table2),
```

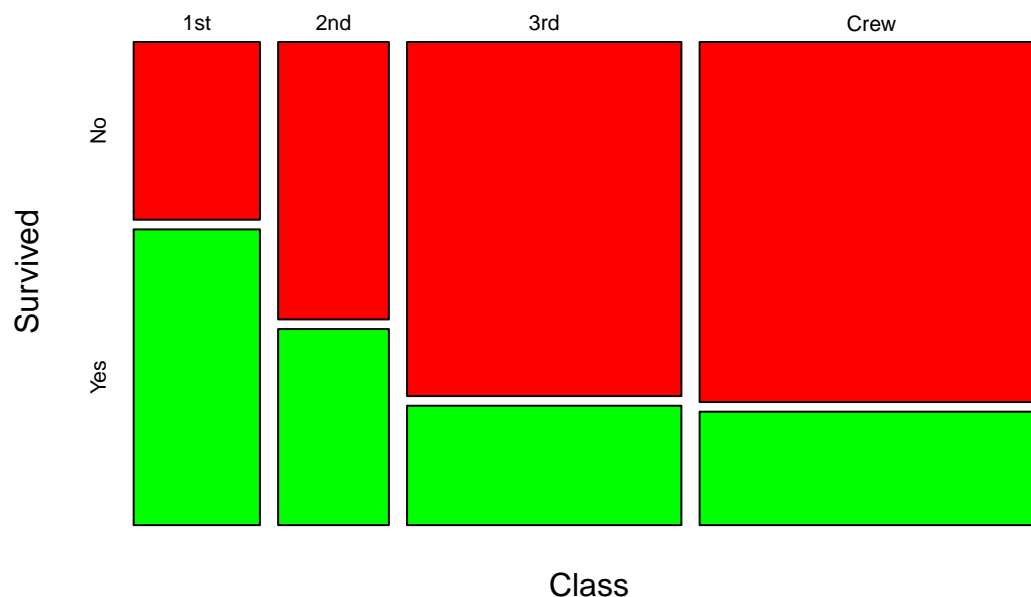


```
main = "Survival by Class (Expanded Titanic Data)",
xlab = "Survived",
ylab = "Number of Passengers")
```



```
# Mosaic plot of Class vs Survived in titanic2
mosaicplot(~ Class + Survived,
  data = titanic2,
  col = c("red", "green"),
  main = "Mosaic Plot of Class vs Survived (Expanded Titanic Data)")
```

Mosaic Plot of Class vs Survived (Expanded Titanic Data)



EXERCISE 7

```
# Load the Airquality dataset
airquality <- read.csv("./datasets/airquality.csv", header = TRUE, sep = ",")

# Clean the dataset
airquality <- na.omit(airquality)

# Calculate the correlation matrix
print(correlation_matrix <- cor(airquality))
```

```
##           Ozone      Solar.R      Wind      Temp      Month
## Ozone      1.000000000  0.34834169 -0.61249658  0.6985414  0.142885168
## Solar.R    0.348341693  1.00000000  -0.12718345  0.2940876 -0.074066683
## Wind      -0.612496576 -0.12718345  1.00000000  -0.4971897 -0.194495804
## Temp       0.698541410  0.29408764 -0.49718972  1.0000000  0.403971709
## Month      0.142885168 -0.07406668 -0.19449580  0.4039717  1.000000000
## Day       -0.005189769 -0.05775380  0.04987102 -0.0965458 -0.009001079
##           Day
## Ozone      -0.005189769
## Solar.R    -0.057753801
## Wind       0.049871017
## Temp      -0.096545800
## Month     -0.009001079
## Day       1.000000000
```

```
# High correlation between Ozone and Temp (0.6985414)
# Low correlation between Wind and Temp (-0.4579883)
```

```
# Load the Cars dataset
cars <- read.csv("./datasets/cars.csv", header = TRUE, sep = ",")
```

```

# Clean the dataset
cars <- na.omit(cars)

# Calculate the correlation matrix
print(correlation_matrix <- cor(cars))

##           X      speed      dist
## X      1.0000000 0.9854590 0.8176576
## speed 0.9854590 1.0000000 0.8068949
## dist  0.8176576 0.8068949 1.0000000

# High correlation between all variables

# Perform a simple random sampling of 50 examples.

# Load the Airquality dataset
airquality <- read.csv("./datasets/airquality.csv", header = TRUE, sep = ",")

# Clean the dataset
airquality <- na.omit(airquality)

# Perform simple random sampling
airquality <- airquality[sample(nrow(airquality), 50), ]
head(airquality)

##      Ozone Solar.R Wind Temp Month Day
## 3       12      149 12.6   74     5   3
## 73      10      264 14.3   73     7  12
## 62     135      269  4.1   84     7   1
## 51      13      137 10.3   76     6  20
## 138     13      112 11.5   71     9  15
## 30     115      223  5.7   79     5  30

# Perform a stratified random sampling of 5 examples each month.
library(dplyr)

# Load the Airquality dataset
airquality <- read.csv("./datasets/airquality.csv", header = TRUE, sep = ",")

# Clean the dataset
airquality <- na.omit(airquality)

# Perform stratified random sampling
airquality <- airquality %>%
  group_by(Month) %>%
  sample_n(5)
head(airquality)

## # A tibble: 6 x 6
## # Groups:   Month [2]
##   Ozone Solar.R Wind Temp Month Day
##   <int>   <int> <dbl> <int> <int> <int>
## 1    11     290   9.2    66     5   13
## 2    23     299   8.6    65     5    7
## 3     4      25   9.7    61     5   23
## 4    34     307  12     66     5   17

```

```
## 5      11      320 16.6    73      5      22
## 6      37      284 20.7    72      6      17
```

EXERCISE 8

```
# Load the dataset
sales <- read.table("./datasets/sales.txt", header = TRUE, sep = "")

# Calculate the total sales per store using Aggregate()
total_sales_per_store <- aggregate(sales_amount ~ store, data = sales, sum)

total_sales_per_store
```

```
##      store sales_amount
## 1 Store_A          2430
## 2 Store_B          2980
## 3 Store_C          1820
```

```
# Load the dataset
sales <- read.table("./datasets/sales.txt", header = TRUE, sep = "")

# Find the avg sales amount for each product category
avg_sales_per_category <- aggregate(sales_amount ~ product_category,
                                   data = sales,
                                   mean)

avg_sales_per_category
```

```
##  product_category sales_amount
## 1      Clothing      660.000
## 2    Electronics     1463.333
## 3      Groceries      430.000
```

```
pdf("./documents/exercice8/ex8_total_sales_per_store_category_matrix.pdf")
```

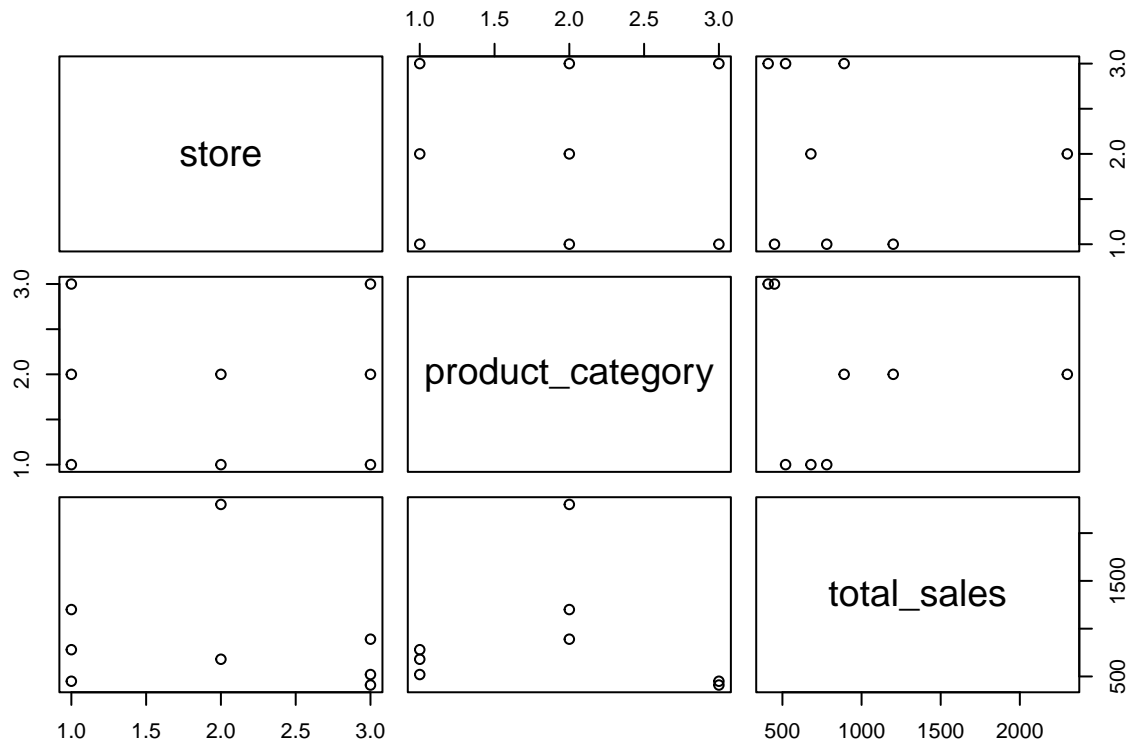
```
library(dplyr)
library(gridExtra)
```

```
# Load the dataset
sales <- read.table("./datasets/sales.txt", header = TRUE, sep = "")
str(sales)
```

```
## 'data.frame':   8 obs. of  3 variables:
## $ store      : chr  "Store_A" "Store_A" "Store_A" "Store_B" ...
## $ product_category: chr  "Electronics" "Groceries" "Clothing" "Electronics" ...
## $ sales_amount  : int  1200 450 780 2300 680 410 520 890
```

```
# Group the data by store and product category, calculate the total sales
total_sales_per_store_category <- sales %>%
  group_by(store, product_category) %>%
  summarise(total_sales = sum(sales_amount), .groups = "drop")

plot(total_sales_per_store_category)
```



```
dev.off()
```

```
## pdf
## 3
```

```
pdf("./documents/exercice8/ex8_total_sales_per_store_category_stacked.pdf")
```

```
library(dplyr)
library(ggplot2)
```

```
# Load the dataset
```

```
sales <- read.table("./datasets/sales.txt", header = TRUE, sep = "")
str(sales)
```

```
## 'data.frame': 8 obs. of 3 variables:
## $ store : chr "Store_A" "Store_A" "Store_A" "Store_B" ...
## $ product_category: chr "Electronics" "Groceries" "Clothing" "Electronics" ...
## $ sales_amount : int 1200 450 780 2300 680 410 520 890
```

```
# Group the data by store and product category, calculate the total sales
```

```
total_sales_per_store_category <- sales %>%
  group_by(store, product_category) %>%
  summarise(total_sales = sum(sales_amount), .groups = "drop")
```

```
# Create the stacked bar plot
```

```
ggplot(total_sales_per_store_category, aes(x = store, y = total_sales, fill = product_category)) +
  geom_bar(stat = "identity") +
  geom_text(aes(label = total_sales,
                position = position_stack(vjust = 0.5),
                size = 3) +
  labs(title = "Total Sales per Store and Product Category",
        x = "Store",
```

```
y = "Total Sales Amount") +
theme_minimal()
```



```
dev.off()
```

```
## pdf
## 3
```

EXERCISE 9

```
library(dplyr)

# Load the datasets
customers <- read.table("./datasets/customers.txt", header = TRUE, sep = "")
orders <- read.table("./datasets/orders.txt", header = TRUE, sep = "")

# Merge the two datasets by customer_id
merged_data <- inner_join(customers, orders, by = "customer_id")

# Count the number of unique customers in the merged dataset
num_unique_customers <- merged_data %>%
  distinct(customer_id) %>%
  n_distinct()

# Print the number of unique customers
print(num_unique_customers)
```

```
## [1] 4
# Count the number of orders placed by each customer
order_counts <- table(merged_data$customer_id)

# Print the number of orders placed by each customer
print(order_counts)

##
## 101 102 103 104
##    2    1    1    1
# Save the merged dataset as a new CSV file called "customer_orders.csv."
write.csv(merged_data, file = "./datasets/customer_orders.csv", row.names = FALSE)
```