

## EXPERIMENT NO : 08

**AIM:** Analyze relationships between multiple variables using visualizations and correlation analysis in R for better data interpretation.

### Definition

- **Correlation** measures the **strength and direction** of the relationship between two numeric variables.
- The correlation coefficient (**r**) ranges between **-1 and 1**:
- **r = 1** → Perfect positive correlation (both variables increase together)
- **r = 0** → No correlation (no relationship between variables)
- **r = -1** → Perfect negative correlation (one variable increases, the other decreases)

### Types of Correlation in R

- **Pearson correlation:** Measures the **linear** relationship between two continuous variables.
- **Spearman correlation:** Measures **monotonic** relationships (useful when data is not normally distributed).
- **Kendall correlation:** Used for **ordinal** data or small datasets.

### Visualizing Relationships Between Multiple Variables

- A. Scatter Plots (Pairwise Relationships)** Used for examining relationships between two numeric variables.
- B. Pair Plot (Visualizing Relationships for Multiple Variables)** A pair plot helps visualize relationships between all numeric variables.
- C. Boxplots (Comparing a Numeric Variable Across Categories)** Useful for checking **distribution differences** across categorical variables.
- D. Bar Plots (Comparing Categorical Variables)** Used to **compare counts or proportions in categorical variables**.

### CODE:

```
library(ggplot2)
library(corrplot)
library(GGally)

# Load the Iris dataset
data(iris)

# 1. Scatter plot of Sepal.Length vs Petal.Length
ggplot(iris, aes(x = Sepal.Length, y = Petal.Length, color = Species)) +
  geom_point() +
  labs(title = "Scatter Plot of Sepal Length vs Petal Length") +
```

```

theme_minimal()

# 2. Pair plot of all numeric variables in the Iris dataset
ggpairs(iris, aes(color = Species))

# 3. Compute and print the correlation matrix
cor_matrix <- cor(iris[, 1:4]) # Using only numeric columns
print(cor_matrix)

# 4. Correlation heatmap
corrplot(cor_matrix, method = "circle", type = "upper",
         title = "Correlation Heatmap", tl.cex = 0.8, tl.col = "black")

# 5. Pearson correlation between Sepal.Length and Petal.Length
pearson_corr <- cor(iris$Sepal.Length, iris$Petal.Length)
print(paste("Pearson Correlation between Sepal.Length and Petal.Length:", pearson_corr))

```

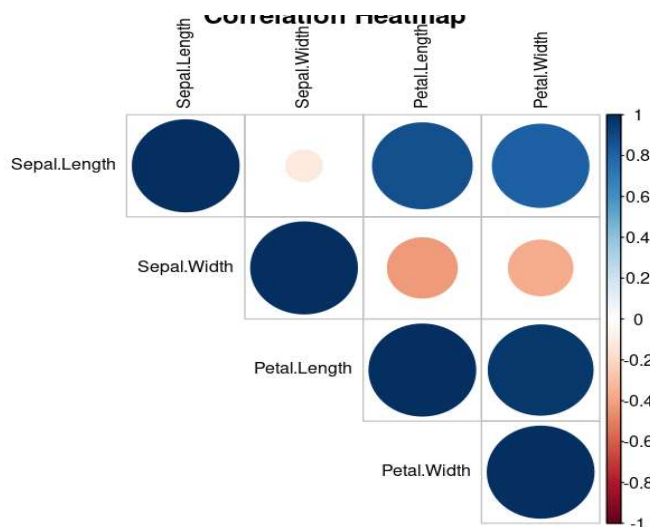
#### OUTPUT:

```

Sepal.Length Sepal.Width Petal.Length Petal.Width
Sepal.Length  1.0000000 -0.1175698  0.8717538  0.8179411
Sepal.Width   -0.1175698  1.0000000 -0.4284401 -0.3661259
Petal.Length   0.8717538 -0.4284401  1.0000000  0.9628654
Petal.Width    0.8179411 -0.3661259  0.9628654  1.0000000

"Pearson Correlation between Sepal.Length and Petal.Length: 0.871753775886583"

```



**CONCLUSION:** Hence we successfully implemented relationships between multiple variables using visualizations and correlation analysis in R for better data interpretation.