

Iris Analysis Report

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Introduction

Analyse du dataset Iris pour un mini projet Data Analyst.

Objectifs : - Statistiques descriptives - Visualisations - Tests statistiques - Régression linéaire

Analyse et visualisations

```
# -----
# Charger le dataset
# -----
data(iris)

# Aperçu
head(iris)

##   Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1         5.1       3.5        1.4       0.2   setosa
## 2         4.9       3.0        1.4       0.2   setosa
## 3         4.7       3.2        1.3       0.2   setosa
## 4         4.6       3.1        1.5       0.2   setosa
## 5         5.0       3.6        1.4       0.2   setosa
## 6         5.4       3.9        1.7       0.4   setosa

summary(iris)

##   Sepal.Length     Sepal.Width    Petal.Length    Petal.Width
##   Min.   :4.300   Min.   :2.000   Min.   :1.000   Min.   :0.100
##   1st Qu.:5.100  1st Qu.:2.800  1st Qu.:1.600  1st Qu.:0.300
##   Median :5.800  Median :3.000  Median :4.350  Median :1.300
##   Mean   :5.843  Mean   :3.057  Mean   :1.758  Mean   :1.199
##   3rd Qu.:6.400  3rd Qu.:3.300  3rd Qu.:5.100  3rd Qu.:1.800
##   Max.   :7.900  Max.   :4.400  Max.   :6.900  Max.   :2.500
##
##   Species
##   setosa      :50
##   versicolor :50
##   virginica  :50
##
## 
## 

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 ...
```

```

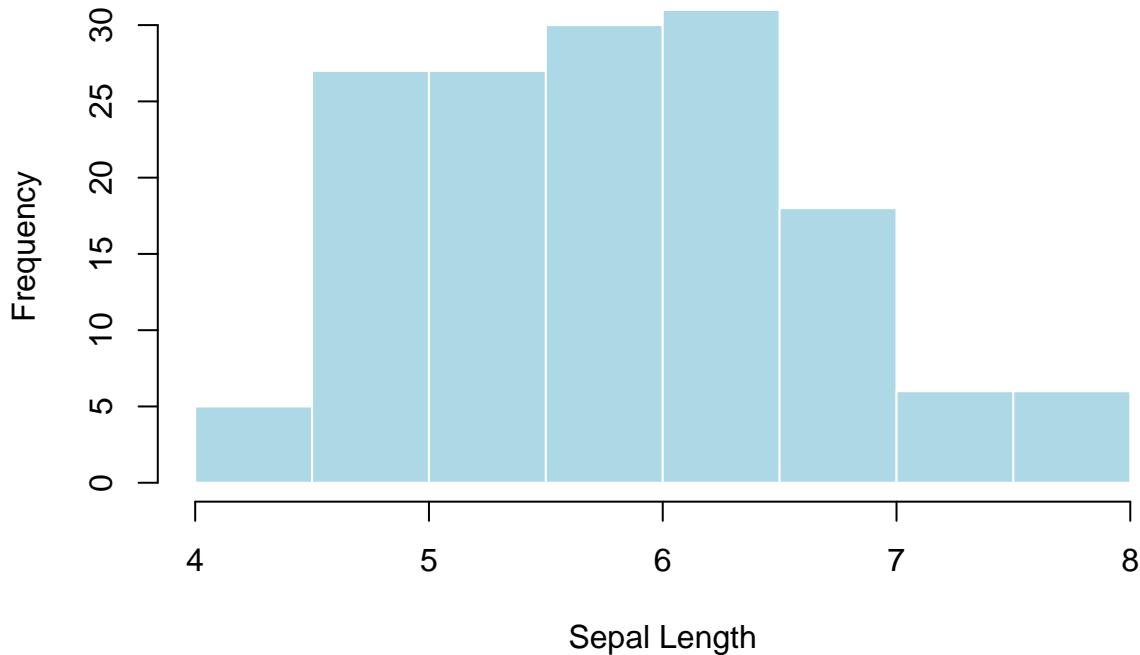
## $ Sepal.Width : num  3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
## $ Petal.Length: num  1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
## $ Petal.Width : num  0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
## $ Species      : Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 ...
# -----
# Visualisations
# -----
```

Histogramme

```

hist(
  iris$Sepal.Length,
  main = "Distribution de Sepal Length",
  xlab = "Sepal Length",
  col = "lightblue",
  border = "white"
)
```

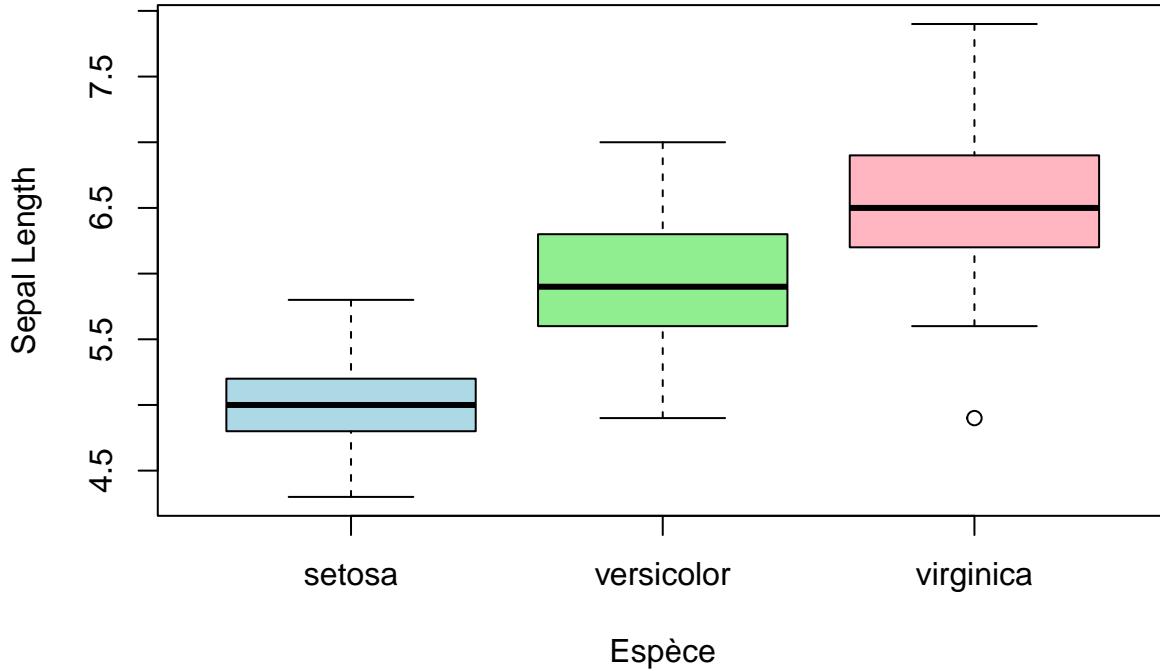
Distribution de Sepal Length



```

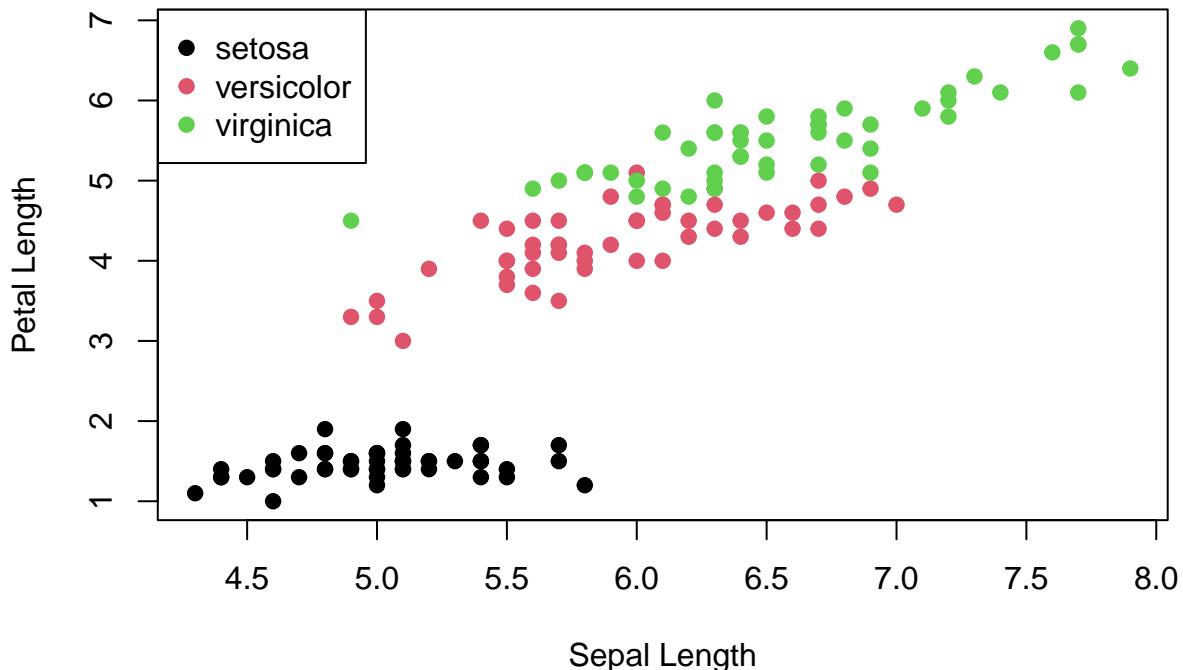
# Boxplot
boxplot(
  Sepal.Length ~ Species,
  data = iris,
  main = "Sepal Length selon l'espèce",
  xlab = "Espèce",
  ylab = "Sepal Length",
  col = c("lightblue", "lightgreen", "lightpink")
)
```

Sepal Length selon l'espèce



```
# Scatter plot
plot(
  iris$Sepal.Length,
  iris$Petal.Length,
  main = "Relation entre Sepal Length et Petal Length",
  xlab = "Sepal Length",
  ylab = "Petal Length",
  pch = 19,
  col = as.numeric(iris$Species)
)
legend(
  "topleft",
  legend = levels(iris$Species),
  col = 1:3,
  pch = 19
)
```

Relation entre Sepal Length et Petal Length



```

# -----
# Test ANOVA et Tukey HSD
# -----
anova_model <- aov(Sepal.Length ~ Species, data = iris)
summary(anova_model)

##           Df Sum Sq Mean Sq F value Pr(>F)
## Species      2   63.21   31.606   119.3 <2e-16 ***
## Residuals  147   38.96    0.265
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
tukey_result <- TukeyHSD(anova_model)
tukey_result

## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = Sepal.Length ~ Species, data = iris)
##
## $Species
##          diff      lwr      upr p adj
## versicolor-setosa  0.930  0.6862273 1.1737727   0
## virginica-setosa  1.582  1.3382273 1.8257727   0
## virginica-versicolor 0.652  0.4082273 0.8957727   0

# Ici, toutes les p-values ajustées = 0,
# donc chaque paire d'espèces présente une différence significative
# de longueur de sépales.

```

```

# -----
# Régression linéaire et plot ggplot2
# -----
linear_model <- lm(Sepal.Length ~ Petal.Length, data = iris)
summary(linear_model)

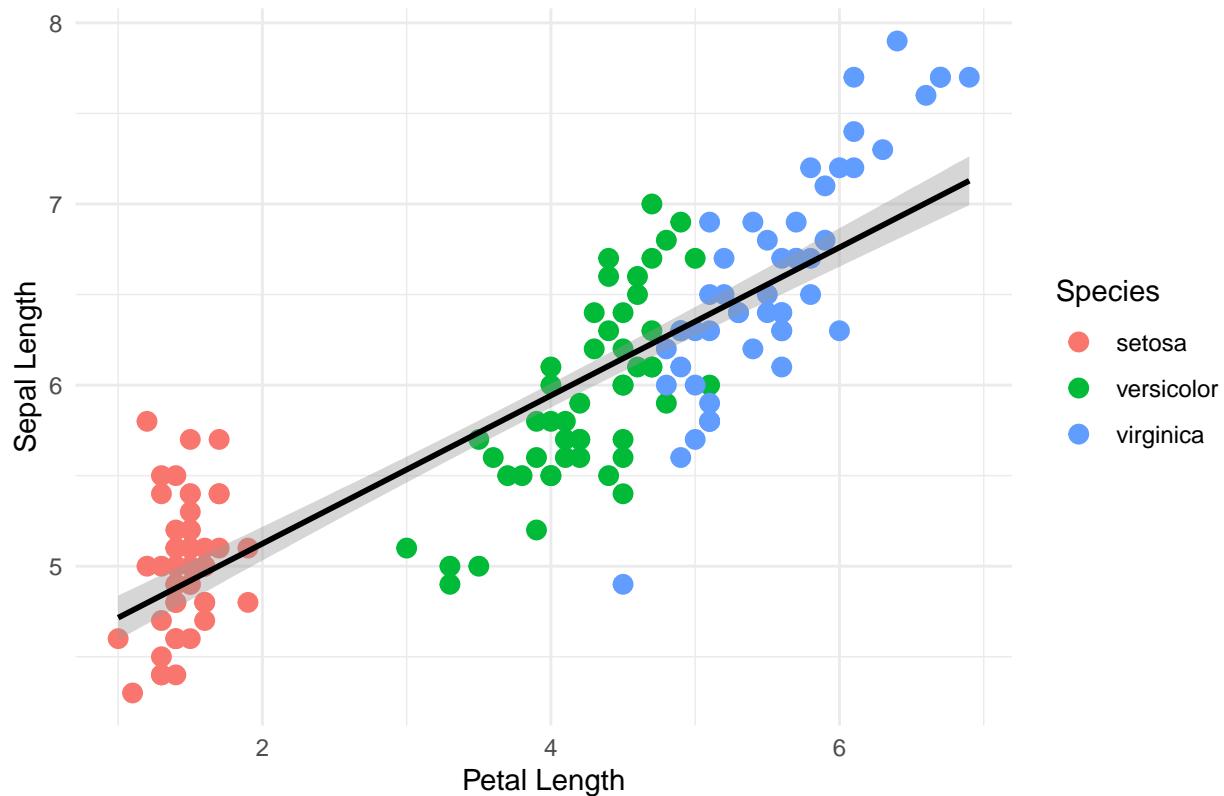
## 
## Call:
## lm(formula = Sepal.Length ~ Petal.Length, data = iris)
## 
## Residuals:
##       Min        1Q    Median        3Q       Max
## -1.24675 -0.29657 -0.01515  0.27676  1.00269
## 
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 4.30660   0.07839  54.94   <2e-16 ***
## Petal.Length 0.40892   0.01889   21.65   <2e-16 ***
## ---    
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Residual standard error: 0.4071 on 148 degrees of freedom
## Multiple R-squared:  0.76, Adjusted R-squared:  0.7583 
## F-statistic: 468.6 on 1 and 148 DF,  p-value: < 2.2e-16

library(ggplot2)
ggplot(iris, aes(x = Petal.Length, y = Sepal.Length)) +
  geom_point(aes(color = Species), size = 3) +
  geom_smooth(method = "lm", se = TRUE, color = "black") +
  labs(
    title = "Régression linéaire : Sepal.Length en fonction de Petal.Length",
    x = "Petal Length",
    y = "Sepal Length"
  ) +
  theme_minimal()

## `geom_smooth()` using formula = 'y ~ x'

```

Régression linéaire : Sepal.Length en fonction de Petal.Length



```
# - Coefficient positif et significatif : Sepal.Length augmente avec Petal.Length
# - R-squared = 0.76 : modèle explique 76% de la variance
```

Conclusion

- Les analyses montrent que chaque espèce d'iris a des sépales de longueur moyenne différente.
- La régression linéaire confirme que Petal.Length est un bon prédicteur de Sepal.Length.