

8)a

9) c logistical regre

```
data <- c(4, 7, 8, 9, 7, 10, 5, 7, 6, 8)
mean_value <- mean(data)
cat("Mean:", mean_value, "\n")
median_value <- median(data)
cat("Median:", median_value, "\n")
mode_value <- as.numeric(names(sort(table(data),
decreasing = TRUE)[1]))
cat("Mode:", mode_value, "\n")
```

```
8)b
input <- mtcars[,c("am","mpg","hp")]
print(head(input))
input <- mtcars
result <- aov(mpg~hp*am,data = input)
print(summary(result))
result <- aov(mpg~hp+am,data = input)
print(summary(result))
result1 <- aov(mpg~hp*am,data = input)
result2 <- aov(mpg~hp+am,data = input)
print(anova(result1,result2))
```

```
9)a Linear regression
height <- c(150, 160, 170, 180, 190)
weight <- c(50, 60, 70, 80, 90)
model <- lm(weight ~ height)
plot(height, weight, main = "Height vs.
Weight",xlab = "Height (cm)", ylab = "Weight
(kg)", pch = 16)
abline(model, col = "red")
```

```
9)b Multiple regresssion
age <- c(25, 30, 35, 40, 45)
height_multi <- c(150, 160, 170, 180, 190)
weight_multi <- c(50, 60, 65, 75, 85)
multiple_model <- lm(weight_multi ~ height_multi
+ age)
summary(multiple_model)
library(ggplot2)
ggplot(data = data.frame(weight_multi, fitted =
fitted(multiple_model)), aes(x = fitted, y =
weight_multi)) +geom_point()
+geom_smooth(method = "lm", color = "blue", se =
FALSE) +ggtitle("Multiple Regression: Residuals
vs Fitted Values") +
theme_minimal()
```

```
age <- c(20, 30, 40, 50, 60)
buy <- c(0, 1, 0, 1, 1)
model <- glm(buy ~ age, family = binomial)
summary(model)
predicted_probs <- predict(model, newdata =
data.frame(age = c(25, 35)), type = "response")
print(predicted_probs)
age_seq <- seq(min(age), max(age), length.out = 100)
fitted_probs <- predict(model, newdata = data.frame(age =
age_seq), type = "response")
plot(age, buy, main = "Logistic Regression: Age vs.
Buying",
xlab = "Age", ylab = "Buying Probability", pch = 16,
col = "blue", ylim = c(0, 1))
lines(age_seq, fitted_probs, col = "red", lwd = 2)
```

9)d poisiion regression

```
input <- warpbreaks
print(head(input))
output <-glm(formula = breaks ~ wool+tension, data =
warpbreaks,
family = poisson)
print(summary(output))
```

```
10)a time series
snowfall <- c(790,1170.8,860.1,1330.6,630.4,9
11.5,683.5,996.6,783.2,982,881.8,1021)
snowfall_timeseries<- ts(snowfall,start =
c(2013,1),frequency = 12)
print(snowfall_timeseries)
png(file = "snowfall.png")
plot(snowfall_timeseries)
dev.off()
```

```
10)b non linear leat square
xvalues <- c(1.6, 2.1, 2, 2.23.19, 2.21)
yvalues <- c(5.19, 7.43, 6.94, 8.11, 18)
png(file = "nls.png")
plot(xvalues, yvalues)
model <- nls(yvalues ~ b1*xvalues^2 + b2, start = list(b1 =
1, b2 = 3))
lines(seq(min(xvalues), max(xvalues), length.out = 100),
predict(model, newdata = data.frame(xvalues =
seq(min(xvalues), max(xvalues), length.out = 100))))
dev.off()
print(sum(resid(model)^2))
print(confint(model))
```

## 10) Decision tree

<pre> data("iris") install.packages("caret") install.packages("C50") library(caret) library(C50) set.seed(7) inTraininglocal &lt;- createDataPartition(iris\$Species, p = .70, list = FALSE) training &lt;- iris[inTraininglocal, ] testing &lt;- iris[-inTraininglocal, ] model &lt;- C5.0(Species ~ ., data = training) summary(model) pred &lt;- predict(model, testing[, -5]) # use predict() instead of predict.C5.0() a &lt;- table(testing\$Species, pred) accuracy &lt;- sum(diag(a)) / sum(a) print(accuracy) plot(model) </pre>	<pre> library("MASS") print(str(Cars93)) car_data&lt;- data.frame(Cars93\$AirBags, Cars93\$Type) car_data = table(Cars93\$AirBags, Cars93\$Type) print(car_data) print(chisq.test(car_data)) </pre>
<p>11) Normal Dist</p> <p><u>dnorm</u></p> <pre> x &lt;- seq(-10, 10, by = .1) y &lt;- dnorm(x, mean = 2.5, sd = 0.5) png(file = "dnorm.png") plot(x,y) dev.off() </pre> <p><u>pnorm</u></p> <pre> x &lt;- seq(-10,10,by = .2) y &lt;- pnorm(x, mean = 2.5, sd = 2) png(file = "pnorm.png") plot(x,y) dev.off() </pre> <p><u>qnorm</u></p> <pre> x &lt;- seq(-10,10,by = .2) y &lt;- pnorm(x, mean = 2.5, sd = 2) png(file = "pnorm.png") plot(x,y) dev.off() </pre> <p><u>rnorm</u></p> <pre> y &lt;- rnorm(50) png(file = "rnorm.png") hist(y, main = "Normal DIstribution") dev.off() </pre> <div data-bbox="600 1384 906 1832"> <pre> x &lt;- seq(0,50,by = 1) y &lt;- dbinom(x,50,0.5)  png(file = "dbinom.png")  plot(x,y)  dev.off()  plot(x,y) </pre> </div>	<p>12) b t test</p> <pre> x &lt;- c(0.593, 0.142, 0.329, 0.691, 0.231, 0.793, 0.519, 0.392, 0.418) t.test(x, alternative="greater", mu=0.3) </pre> <p>12) c t test</p> <pre> install.packages("randomForest") library(party) print(head(readingSkills)) library(party) library(randomForest) output.forest &lt;- randomForest(nativeSpeaker ~ age + shoeSize + score, data = readingSkills) print(output.forest) </pre> <p>Week 2 associate</p> <p>Step 1 : load the data set ( open weka and click on ‘Explore’, click open file and load the contact lenses arff dataset)</p> <p>Step 2 select the aprori alg (go to ‘associate’ tab</p> <p>, clic ‘ choose ‘ and click weka association apriori from the list) . step 3 configure and run apriori (click on the ‘associator’ and set minisupped and minmetric, click start to run alg)</p>