## PRACTICAL 2 R

## BDATC01/1717/2022

### 2024-11-14

```
#1.Define the coefficients of the polynomial
#x^5+3x^4+((4x^2)/5)+3x+1
coefficients <- c(1, 3, 0, 4/5, 3, 1) # Coefficients for x^5, x^4, x^3, x^2, x, constant term
# Find the roots using the polyroot function
roots <- polyroot(coefficients)</pre>
# Print the roots
print(roots)
## [1] -0.3344753+0.0000000i -1.2588390-0.0000000i 0.5565643-0.7954688i
## [4] 0.5565643+0.7954688i -2.5198143+0.0000000i
library(rootSolve)
# Define the function
f<-function(x) {
  x^3 + 4*x^2 + x + 6
# Define the first derivative
f_prime <-function(x) {</pre>
  3*x^2 + 8*x + 1
# Define the second derivative
f_double_prime <-function(x) {</pre>
  6*x + 8
# Find the roots of the first derivative
stationary_points <-uniroot.all(f_prime, c(-10, 10))
# Determine the nature of each stationary point
nature_of_points <-sapply(stationary_points,</pre>
                           function(x) {
                             if (f_double_prime(x) > 0) {
                               return("Minimum")
                             } else if (f_double_prime(x) < 0) {</pre>
                               return("Maximum")
                             } else {
                               return("Inflection")
                           })
# Output the results
stationary_points
```

## [1] -2.5350850 -0.1314719

```
nature_of_points
## [1] "Maximum" "Minimum"
# Load necessary library
library(Deriv)
library(pracma)
## Attaching package: 'pracma'
## The following objects are masked from 'package:rootSolve':
##
       gradient, hessian
cot <- function(x) 1 / tan(x)</pre>
# Define the function f(x)
f <- function(x) {</pre>
  2*x * \cot(x) + x * \sin(x^2) + \exp(3*x^3) + x^2 + 4
}
\# i) Differentiate the function with respect to x
# Use the Deriv package to differentiate the function
library(Deriv)
f_prime <- Deriv(f, "x")</pre>
print(f_prime)
## function (x)
## {
       .e1 <- x^2
##
##
       2 * \cot(x) + \sin(.e1) + x * (2 + x * (2 * \cos(.e1) + 9 *
##
           \exp(3 * x^3) - 2/(\cos(x)^2 * \tan(x)^2)
## }
# ii) Integrate the function over the range 3 < x < 6
# Use integrate function for numerical integration
integral_result <- integrate(f, lower = 3, upper = 6)</pre>
print(integral_result$value)
## [1] 8.179504e+278
#Binomial distribution
# Define parameters for the binomial distribution
n<- 6
             # Number of trials in the binomial distribution
                # Probability of success in each trial
p < -0.6
# Sample sizes for the simulations
sample_sizes \leftarrow c(10, 100, 1000, 10000)
# Create a 2x2 plotting grid to display all the histograms on the same plot
par(mfrow=c(2, 2))
# Loop over each sample size
for (sample_size in sample_sizes) {
```

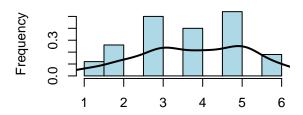
```
# Simulate binomial observations
  data <- rbinom(sample_size, n, p)</pre>
  # Plot histogram for the current sample size
  hist(data,
       main = paste("Sample Size =", sample_size),
                                                      # Title with sample size
       xlab = "Number of Successes",
                                                       # X-axis label
       ylab = "Frequency",
                                                       # Y-axis label
       col = "lightblue",
                                                      # Histogram color
       border = "black",
                                                      # Border color
       probability = TRUE,
                                                      # Normalize the histogram to show density
       breaks = 10)
                                                      # Number of bins
  # Add density line to the histogram
  lines(density(data), col = "black", lwd = 2)
                                                      # Density line in red with thicker line width
}
```

## Sample Size = 10

## Fredunds 3.0 3.5 4.0 4.5 5.0

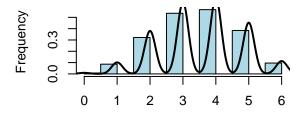
Number of Successes

## Sample Size = 100



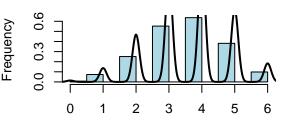
Number of Successes

## Sample Size = 1000



Number of Successes

## Sample Size = 10000



Number of Successes

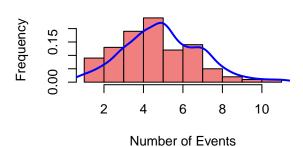
```
#Poisson distribution
# Simulate data from a Poisson distribution with lambda = 5 for different sample sizes
p1 <- rpois(n = 10, lambda = 5)  # Sample of 10 from Poisson distribution
p2 <- rpois(n = 100, lambda = 5)  # Sample of 100 from Poisson distribution
p3 <- rpois(n = 1000, lambda = 5)  # Sample of 1000 from Poisson distribution
p4 <- rpois(n = 10000, lambda = 5)  # Sample of 10000 from Poisson distribution
# Create a 2x2 plotting grid to display all the histograms on the same plot
par(mfrow=c(2,2))</pre>
```

```
# Plot for sample size = 10
hist(p1, main = "Sample size 10",
     xlab = "Number of Events", ylab = "Frequency",
     col = "lightcoral", border = "black",
     probability = TRUE, breaks = 5)
lines(density(p1), col = "blue", lwd = 2) # Density line in blue
\# Comment: "For n = 10, the histogram is very sparse, with many zero counts. The density line is rough
# Plot for sample size = 100
hist(p2, main = "Sample size 100",
     xlab = "Number of Events", ylab = "Frequency",
     col = "lightcoral", border = "black",
     probability = TRUE, breaks = 10)
lines(density(p2), col = "blue", lwd = 2) # Density line in blue
# Comment: "At n = 100, the histogram starts to resemble the Poisson distribution more closely. The den
# Plot for sample size = 1000
hist(p3, main = "Sample size 1000",
     xlab = "Number of Events", ylab = "Frequency",
     col = "lightcoral", border = "black",
     probability = TRUE, breaks = 15)
lines(density(p3), col = "blue", lwd = 2) # Density line in blue
\# Comment: "With n = 1000, the histogram becomes much smoother and is more representative of the Poisso
# Plot for sample size = 10000
hist(p4, main = "Sample size 10000",
     xlab = "Number of Events", ylab = "Frequency",
     col = "lightcoral", border = "black",
     probability = TRUE, breaks = 20)
lines(density(p4), col = "blue", lwd = 2) # Density line in blue
```

## Sample size 10

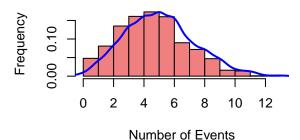
## Freduency 0.0 0.3 0.6 4 5 6 7 8 9

## Sample size 100

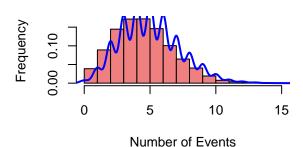


## Sample size 1000

Number of Events



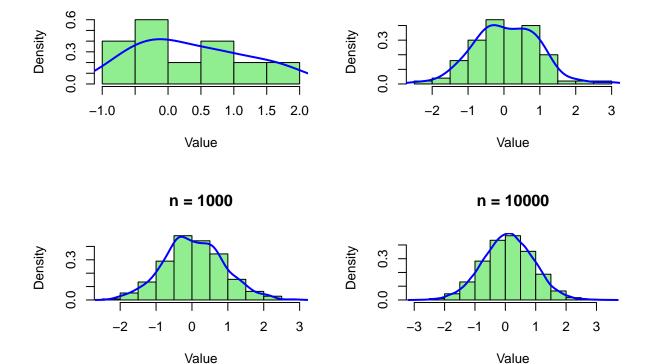
## Sample size 10000



# Comment: "At n = 10000, the histogram is very smooth and closely approximates the theoretical Poisson

```
#Normal distribution
# Simulate data from a normal distribution with mean = 0.1 and sd = 0.84
norm \leftarrow rnorm(10, mean = 0.1, sd = 0.84)
                                               # Sample size = 10
norm1 \leftarrow rnorm(100, mean = 0.1, sd = 0.84)
                                                # Sample size = 100
norm2 \leftarrow rnorm(1000, mean = 0.1, sd = 0.84)
                                                # Sample size = 1000
norm3 \leftarrow rnorm(10000, mean = 0.1, sd = 0.84)
                                               # Sample size = 10000
# Create a 2x2 plotting grid to display all the histograms on the same plot
par(mfrow=c(2,2))
# Plot for sample size = 10
hist(norm, probability = TRUE, main = "n = 10",
     xlab = "Value", ylab = "Density",
     col = "lightgreen", border = "black",
     breaks = 5)
lines(density(norm), col = "blue", lwd = 2) # Density line in blue
# Comment: "For n = 10, the histogram is very jagged and sparse, with a limited representation of the n
# Plot for sample size = 100
hist(norm1, probability = TRUE, main = "n = 100",
     xlab = "Value", ylab = "Density",
     col = "lightgreen", border = "black",
     breaks = 10)
```

n = 100



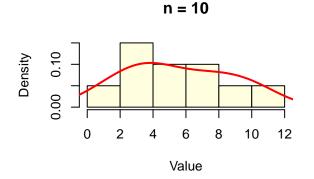
#Chi-square distribution #Simulate data from a chi-square distribution with df = 5

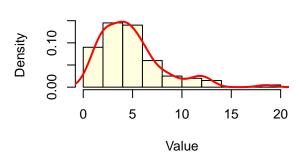
# Comment: "At n = 10000, the histogram is very smooth, and the density line fits the data perfectly. T

ha <- rchisq(10, df = 5) # Sample size = 10 hb <- rchisq(100, df = 5) # Sample size = 100

n = 10

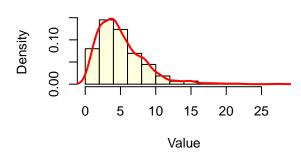
```
hc <- rchisq(1000, df = 5) # Sample size = 1000
hd <- rchisq(10000, df = 5) # Sample size = 10000
# Create a 2x2 plotting grid to display all the histograms on the same plot
par(mfrow=c(2,2))
# Plot for sample size = 10
hist(ha, probability = TRUE, main = "n = 10",
     xlab = "Value", ylab = "Density",
     col = "lightyellow", border = "black",
     breaks = 5)
lines(density(ha), col = "red", lwd = 2) # Density line in red
# Comment: "For n = 10, the histogram is very irregular and does not resemble the typical Chi-square di
# Plot for sample size = 100
hist(hb, probability = TRUE, main = "n = 100",
     xlab = "Value", ylab = "Density",
     col = "lightyellow", border = "black",
     breaks = 10)
lines(density(hb), col = "red", lwd = 2) # Density line in red
# Comment: "At n = 100, the histogram begins to take a more recognizable shape of the Chi-square distri
# Plot for sample size = 1000
hist(hc, probability = TRUE, main = "n = 1000",
     xlab = "Value", ylab = "Density",
     col = "lightyellow", border = "black",
     breaks = 15)
lines(density(hc), col = "red", lwd = 2) # Density line in red
# Comment: "With n = 1000, the histogram closely matches the expected Chi-square distribution. The dens
# Plot for sample size = 10000
hist(hd, probability = TRUE, main = "n = 10000",
     xlab = "Value", ylab = "Density",
     col = "lightyellow", border = "black",
     breaks = 20)
lines(density(hd), col = "red", lwd = 2) # Density line in red
```



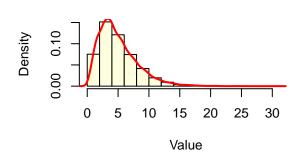


n = 100

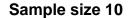
n = 10000



n = 1000

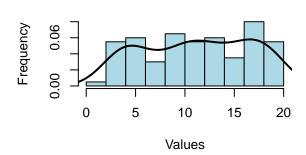


# Comment: "For n = 10000, the histogram is very smooth, and the density line fits the data perfectly. #uniform distribution # Sample sizes for the simulations hh <- runif(10, min = 1, max = 20) # Sample of 10 from Uniform distribution  $hh1 \leftarrow runif(100, min = 1, max = 20)$ # Sample of 100 from Uniform distribution hh2 <- runif(1000, min = 1, max = 20) # Sample of 1000 from Uniform distribution hh3 <- runif(10000, min = 1, max = 20) # Sample of 10000 from Uniform distribution # Create a 2x2 plotting grid to display all the histograms on the same plot par(mfrow=c(2,2)) # Plot for sample size = 10 hist(hh, main = "Sample size 10", xlab = "Values", ylab = "Frequency", col = "lightblue", border = "black", probability = TRUE, breaks = 5) lines(density(hh), col = "black", lwd = 2) # Density line in blue # Comment: "For n = 10, the histogram is discrete with some irregularity. The density line is rough, re # Plot for sample size = 100 hist(hh1, main = "Sample size 100", xlab = "Values", ylab = "Frequency", col = "lightblue", border = "black", probability = TRUE, breaks = 10)

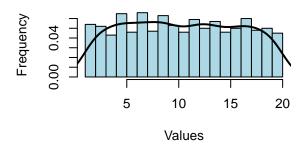


# Programmer of the second of th

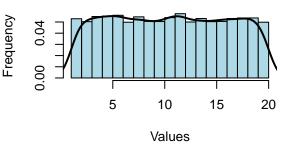
## Sample size 100



## Sample size 1000



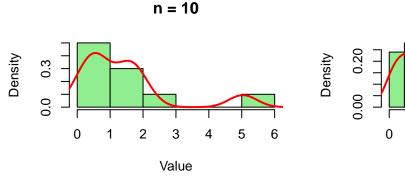
## Sample size 10000

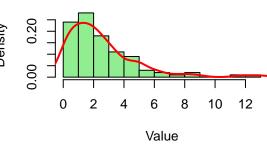


# Comment: "At n = 10000, the histogram is very close to a uniform distribution. The density line is sm

```
#Exponential distribution
# Simulate data from an exponential distribution with rate = 0.45
exp1 <- rexp(10, rate = 0.45)  # Sample size = 10</pre>
```

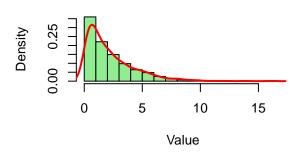
```
exp2 <- rexp(100, rate = 0.45) # Sample size = 100
exp3 <- rexp(1000, rate = 0.45) # Sample size = 1000
exp4 <- rexp(10000, rate = 0.45) # Sample size = 10000
# Create a 2x2 plotting grid to display all the histograms on the same plot
par(mfrow=c(2,2))
# Plot for sample size = 10
hist(exp1, probability = TRUE, main = "n = 10",
     xlab = "Value", ylab = "Density",
     col = "lightgreen", border = "black",
     breaks = 5)
lines(density(exp1), col = "red", lwd = 2) # Density line in red
# Comment: "For n = 10, the histogram is highly irregular and does not closely follow the expected expo
# Plot for sample size = 100
hist(exp2, probability = TRUE, main = "n = 100",
     xlab = "Value", ylab = "Density",
     col = "lightgreen", border = "black",
     breaks = 10)
lines(density(exp2), col = "red", lwd = 2) # Density line in red
# Comment: "At n = 100, the histogram starts to approximate the expected exponential distribution. The
# Plot for sample size = 1000
hist(exp3, probability = TRUE, main = "n = 1000",
     xlab = "Value", ylab = "Density",
     col = "lightgreen", border = "black",
     breaks = 15)
lines(density(exp3), col = "red", lwd = 2) # Density line in red
# Comment: "With n = 1000, the histogram more closely matches the shape of the exponential distribution
# Plot for sample size = 10000
hist(exp4, probability = TRUE, main = "n = 10000",
     xlab = "Value", ylab = "Density",
     col = "lightgreen", border = "black",
     breaks = 20)
lines(density(exp4), col = "red", lwd = 2) # Density line in red
```



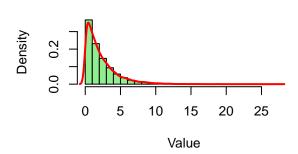


n = 100

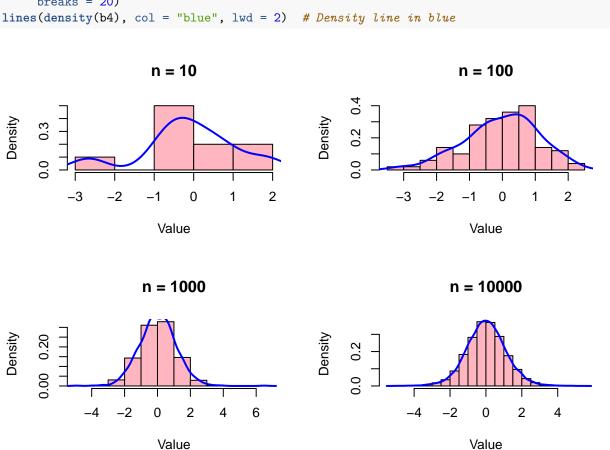
n = 10000



n = 1000



```
\# Comment: "For n = 10000, the histogram is smooth, and the density line fits the data perfectly. This
#t-distribution
# Simulate data from a t-distribution with 10 degrees of freedom
b1 <- rt(10, df = 10)
                           # Sample size = 10
b2 \leftarrow rt(100, df = 10)
                           # Sample size = 100
b3 <- rt(1000, df = 10)
                           # Sample size = 1000
b4 \leftarrow rt(10000, df = 10)
                           # Sample size = 10000
# Create a 2x2 plotting grid to display all the histograms on the same plot
par(mfrow=c(2,2))
# Plot for sample size = 10
hist(b1, probability = TRUE, main = "n = 10",
     xlab = "Value", ylab = "Density",
     col = "lightpink", border = "black",
     breaks = 5)
lines(density(b1), col = "blue", lwd = 2) # Density line in blue
# Comment: "For n = 10, the histogram is very jagged and has heavy tails. The density line is not a goo
# Plot for sample size = 100
hist(b2, probability = TRUE, main = "n = 100",
     xlab = "Value", ylab = "Density",
     col = "lightpink", border = "black",
     breaks = 10)
```



# Comment: "At n = 10000, the histogram closely matches the true t-distribution. The density line fits