# HW8

#### 2024 - 11 - 06

## $\mathbf{Q}\mathbf{1}$

Write a function which takes 2 arguments n and k which are positive integers. It should return the nXn matrix:

$$\begin{pmatrix} k & 1 & 0 & 0 & \dots & 0 & 0 \\ 1 & k & 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & k & 1 & \dots & 0 & 0 \\ 0 & 0 & 1 & k & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \dots & k & 1 \\ 0 & 0 & 0 & 0 & \dots & 1 & k \end{pmatrix}$$

Call the function defined above for n = 6 and k = 5, and provide the matrix you obtain.

One with a for loop, one without.

```
create_matrix <- function(n, k) {
  mat <- matrix(0, n, n)

diag(mat) <- k

for (i in 1:(n-1)) {
  mat[i, i+1] <- 1
  mat[i+1, i] <- 1
}

return(mat)
}

matrix_result <- create_matrix(6, 5)
matrix_result</pre>
```

```
[,1] [,2] [,3] [,4] [,5] [,6]
##
## [1,]
## [2,]
            1
                             0
                                  0
                                        0
                 5
                       1
## [3,]
            0
                 1
                       5
                             1
                                  0
                                        0
## [4,]
                             5
                                        0
            0
                 0
                       1
                                  1
## [5,]
            0
                 0
                       0
                             1
                                  5
                                        1
## [6,]
                       0
                             0
                                  1
                                        5
```

```
create_matrix <- function(n, k) {
  mat <- matrix(0, n, n)

diag(mat) <- k

mat[row(mat) == col(mat) + 1] <- 1
  mat[row(mat) == col(mat) - 1] <- 1

return(mat)
}

matrix_result <- create_matrix(6, 5)
matrix_result</pre>
```

```
## [,1] [,2] [,3] [,4] [,5] [,6]

## [1,] 5 1 0 0 0 0

## [2,] 1 5 1 0 0 0

## [3,] 0 1 5 1 0 0

## [4,] 0 0 1 5 1 0

## [5,] 0 0 0 1 5 1

## [6,] 0 0 0 0 1 5
```

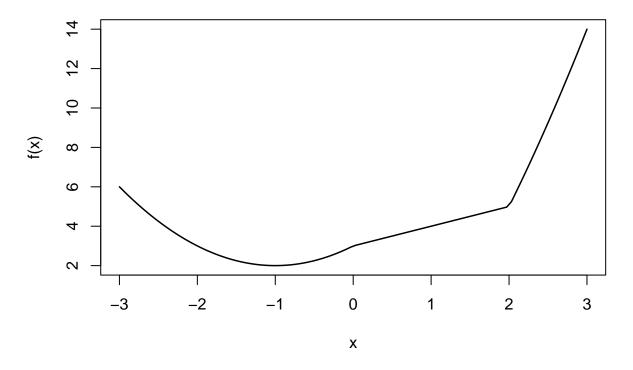
Consider the continuous function

$$f(x) = \begin{cases} x^2 + 2x + 3 & \text{if } x < 0\\ x + 3 & \text{if } 0 \le x < 2\\ x^2 + 4x - 7 & \text{if } 2 \le x \end{cases}$$

Write a function tmpFn which takes a single argument xVec. The function should return the vector of values of the function f(x) evaluated at the values xVec. Plot the function f(x) for -3 < x < 3.

```
tmpFn <- function(xVec) {</pre>
  result <- numeric(length(xVec))</pre>
  result[xVec < 0] <- xVec[xVec < 0]^2 + 2 * xVec[xVec < 0] + 3
  result[xVec \geq 0 \& xVec < 2] <- xVec[xVec \geq 0 \& xVec < 2] + 3
  result[xVec \geq 2] <- xVec[xVec \geq 2]^2 + 4 * xVec[xVec \geq 2] - 7
  return(result)
}
xValues \leftarrow seq(from = -3,
               to = 3,
                length.out = 100)
yValues <- tmpFn(xValues)
plot(x = xValues, y = yValues,
     type = "1",
     col = "black",
     lwd = 1.5,
     xlab = "x",
     ylab = "f(x)",
     main = "Plot of f(x) for -3 < x < 3")
```

Plot of f(x) for -3 < x < 3



### $\mathbf{Q3}$

Greatest common divisor of two integers The greatest common divisor (gcd) of two integers m and n can be calculated using Euclid's Algorithm: Divide m by n. If the remainder is zero, the gcd is n. If not, divide n by the remainder. If the remainder is zero, then the previous remainder is the gcd. If not, continue dividing the remainder into previous remainder until a remainder of zero is obtained. The gcd is the value of the last nonzero remainder. Write a function gcd(m,n) using a while loop to find the gcd of two integers m and n.

```
gcd <- function(m, n) {
    m <- abs(m)
    n <- abs(n)

while (n != 0) {
    remainder <- m %% n
    m <- n
    n <- remainder
}

return(m)
}

# Testing examples
gcd(4*7, 4*4*4*7)

## [1] 28
gcd(47, 93)</pre>
```

### $\mathbf{Q4}$

eQTL mapping. (The following problem was suggested by Professor Dan Nettleton.) Write a function order.matrix which takes in a matrix x and returns a matrix containing the row and column indices of the sorted values of x. Test this function on a 4X3 matrix of independent  $\chi_1^2$  pseudo-random deviates.

```
order.matrix <- function(x) {</pre>
  sorted_indices <- order(x)</pre>
  row_indices <- (sorted_indices - 1) %% nrow(x) + 1</pre>
  col_indices <- (sorted_indices - 1) %/% nrow(x) + 1</pre>
  result <- cbind(row_indices, col_indices)</pre>
  return(result)
set.seed(37)
test_matrix <- matrix(rchisq(4 * 3, df = 1), nrow = 4, ncol = 3)</pre>
dim(test_matrix)
## [1] 4 3
test_matrix
##
               [,1]
                           [,2]
                                       [,3]
## [1,] 0.84692934 0.05563524 2.57108739
## [2,] 1.18006741 1.86200588 0.01183871
## [3,] 0.34755917 2.01569730 0.07067748
## [4,] 0.06079836 2.13410150 0.01434852
indices <- order.matrix(test_matrix)</pre>
indices
##
         row_indices col_indices
    [1,]
##
##
  [2,]
                    4
                                 3
                                 2
## [3,]
                    1
## [4,]
                    4
                                 1
   [5,]
                    3
##
                                 3
## [6,]
                    3
                                 1
##
  [7,]
                    1
                                 1
## [8,]
                    2
                                 1
                    2
                                 2
## [9,]
## [10,]
                    3
                                 2
## [11,]
                                 2
## [12,]
                                 3
dim(indices)
```

```
## [1] 12 2
```

### $Q_5$

Polar representation of a number. Let  $x \in \mathbb{R}^p$ . The polar respresentation of  $\mathbf{x} = (x_1, x_2, ..., x_p)$  is given by  $(R, \theta_1, \theta_2, ..., \theta_{p-1})$ , where:

$$x_1 = R \cos \theta_1$$

$$x_2 = R \sin \theta_1 \cos \theta_2$$

$$x_3 = R \sin \theta_1 \sin \theta_2 \cos \theta_3$$

$$\dots = \dots$$

$$x_{p-1} = R \prod_{i=1}^{p-2} \sin \theta_i \cos \theta_{p-1}$$

$$x_p = R \prod_{i=1}^{p-1} \sin \theta_i,$$

where  $0 \le R < \infty, 0 \le \theta_1 < 2\pi$  and  $0 \le \theta_i < \pi$  i = 2, 3, . . . , p - 1.

(a)

Write a function polaroid which takes in an arbitrary p-dimensional vector  $\mathbf{x}$  and provides its polar representation as a vector, with the first element as R and the remainder being  $\theta_1, \theta_2, \dots, \theta_{p-1}$ .

```
polaroid <- function(x) {</pre>
  R \leftarrow sqrt(sum(x^2))
  p <- length(x)
  theta <- numeric(p - 1)
  if (R > 0) {
    theta[1] \leftarrow atan2(x[2], x[1])
    if (theta[1] < 0) {</pre>
       theta[1] <- theta[1] + 2 * pi
    }
    for (i in 2:(p - 1)) {
      numerator <- sqrt(sum(x[(i + 1):p]^2))</pre>
      denominator <- sqrt(sum(x[i:p]^2))</pre>
      theta[i] <- acos(numerator / denominator)</pre>
    }
  }
  return(c(R, theta))
x \leftarrow c(1, 2, 3)
polar_representation <- polaroid(x)</pre>
polar_representation
```

## [1] 3.7416574 1.1071487 0.5880026

```
polaroid <- function(x) {</pre>
  R \leftarrow sqrt(sum(x^2))
  if (R == 0) return(c(R))
  theta1 \leftarrow atan2(x[2], x[1])
  if (theta1 < 0) {</pre>
    theta1 <- theta1 + 2 * pi
  }
  p <- length(x)
  if (p > 2) {
    numerators <- sqrt(rev(cumsum(rev(x[-c(1, 2)]^2))))</pre>
    denominators <- sqrt(cumsum(rev(x[-1]^2)))</pre>
    theta_rest <- acos(numerators / denominators)</pre>
  } else {
    theta_rest <- numeric(0)</pre>
  return(c(R, theta1, theta_rest))
x \leftarrow c(1, 2, 3)
polar_representation <- polaroid(x)</pre>
print(polar_representation)
```

## [1] 3.7416574 1.1071487 0.0000000 0.5880026

(b)

Write a function normalize which takes in a matrix and returns its normalized form: i.e., the matrix with rows scaled such that the sum of squares of each row is equal to 1.

```
normalize <- function(x) {
  row_norms <- sqrt(rowSums(x^2))

normalized_matrix <- x / row_norms

normalized_matrix[is.nan(normalized_matrix)] <- 0

return(normalized_matrix)
}</pre>
```

```
set.seed(37)
test_matrix <- matrix(rnorm(12), nrow = 4, ncol = 3)
test_matrix</pre>
```

```
## [,1] [,2] [,3]

## [1,] 0.1247540 -0.8283492 0.85595441

## [2,] 0.3820746 -0.3327136 0.21599549

## [3,] 0.5792428 -0.1921595 -0.37770210

## [4,] -0.2937481 1.3629827 0.03869354
```

```
rowSums(test_matrix^2)
## [1] 1.4343838 0.3033334 0.5151063 1.9455071
normalized_matrix <- normalize(test_matrix)</pre>
normalized_matrix
##
              [,1]
                         [,2]
                                     [,3]
## [1,] 0.1041650 -0.6916410 0.7146904
## [2,] 0.6937261 -0.6041022 0.3921792
## [3,] 0.8070718 -0.2677401 -0.5262607
## [4,] -0.2106002 0.9771786 0.0277410
rowSums(normalized_matrix^2)
## [1] 1 1 1 1
```

(c)

Obtain a 1000X5 matrix y of N(0,1) pseudo-random deviates. Use apply and normalize to obtain the normalized values. Call this matrix **z**. We test whether the columns of **z** are uniform on U(-1,1). One may test whether a sample  $x \sim U(-1,1)$  using ks.test(x, "punif", min=-1, max=1) where punif represents the cumulative distribution function of the uniform over range (-1,1). Summarize your results.

```
set.seed(37)
y <- matrix(rnorm(1000 * 5), nrow = 1000, ncol = 5)
normalize <- function(x) {</pre>
  row_norms <- sqrt(rowSums(x^2))</pre>
  normalized_matrix <- sweep(x, 1, row_norms, FUN = "/")</pre>
  normalized_matrix[is.nan(normalized_matrix)] <- 0</pre>
  return(normalized_matrix)
}
z \leftarrow apply(X = y,
            MARGIN = 2
            FUN = function(col) normalize(matrix(col, ncol = 1))
            )
ks_{tests} \leftarrow apply(X = z,
                    MARGIN = 2,
                    FUN = function(column) ks.test(column, "punif", min = -1, max = 1)
                    )
```

```
## Warning in ks.test.default(column, "punif", min = -1, max = 1): ties should not
## be present for the one-sample Kolmogorov-Smirnov test
## Warning in ks.test.default(column, "punif", min = -1, max = 1): ties should not
## be present for the one-sample Kolmogorov-Smirnov test
## Warning in ks.test.default(column, "punif", min = -1, max = 1): ties should not
```

```
## be present for the one-sample Kolmogorov-Smirnov test
## Warning in ks.test.default(column, "punif", min = -1, max = 1): ties should not
## be present for the one-sample Kolmogorov-Smirnov test
## Warning in ks.test.default(column, "punif", min = -1, max = 1): ties should not
## be present for the one-sample Kolmogorov-Smirnov test
ks_tests
## [[1]]
##
   Asymptotic one-sample Kolmogorov-Smirnov test
## data: column
## D = 0.515, p-value < 2.2e-16
## alternative hypothesis: two-sided
##
##
## [[2]]
##
   Asymptotic one-sample Kolmogorov-Smirnov test
##
## data: column
## D = 0.531, p-value < 2.2e-16
## alternative hypothesis: two-sided
##
##
  [[3]]
##
   Asymptotic one-sample Kolmogorov-Smirnov test
##
##
## data: column
## D = 0.525, p-value < 2.2e-16
## alternative hypothesis: two-sided
##
##
## [[4]]
##
##
   Asymptotic one-sample Kolmogorov-Smirnov test
##
## data: column
## D = 0.504, p-value < 2.2e-16
## alternative hypothesis: two-sided
##
##
  [[5]]
##
   Asymptotic one-sample Kolmogorov-Smirnov test
## data: column
## D = 0.51, p-value < 2.2e-16
```

The KS test results strongly suggest that the columns of the normalized matrix z are not uniformly distributed on (-1,1). The deviations are significant, implying that the normalization did not transform the data into a

## alternative hypothesis: two-sided

uniform distribution over this range. Further investigation or a different transformation might be needed if the goal is to achieve a uniform distribution.

#### (d)

Obtain polar representations of **y** using your function polaroid and test whether  $R^2 \sim \chi_5^2$  distribution. Provide a page of histograms or boxplots of  $\theta_1, \theta_2, \theta_3, \theta_4$ . Test whether these are from the uniform distributions on their respective ranges, i.e.,  $[0, 2\pi)$  for  $\theta_1$  and  $[0, \pi)$  for  $\theta_2, \theta_3, \theta_4$ .

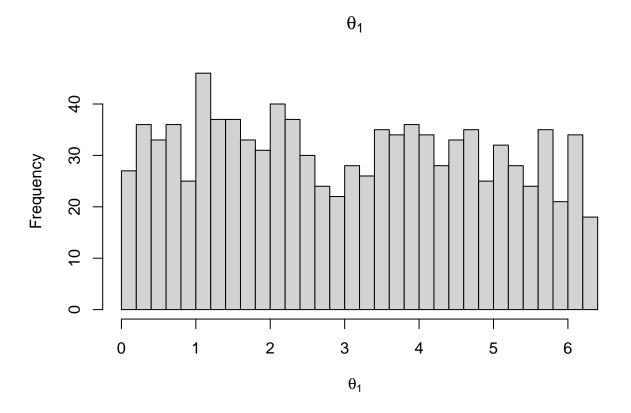
```
set.seed(37)
polaroid <- function(x) {</pre>
  # Calculate R
  R \leftarrow sqrt(sum(x^2))
  # If R is zero, return R with no angles (undefined case)
  if (R == 0) return(c(R))
  # Calculate the first angle theta1 using atan2
  theta1 \leftarrow atan2(x[2], x[1])
  if (theta1 < 0) theta1 <- theta1 + 2 * pi</pre>
  p <- length(x)</pre>
  # Calculate the remaining angles, only if p > 2
  if (p > 2) {
    numerators <- sqrt(rev(cumsum(rev(x[-c(1, 2)]^2))))</pre>
    denominators <- sqrt(cumsum(rev(x[-1]^2))[-1])</pre>
    # Ensure lengths match for division
    if (length(numerators) != length(denominators)) {
      stop("Numerators and denominators lengths do not match.")
    }
    # Calculate theta_rest and clamp values to be in [-1, 1]
    cos_values <- numerators / denominators</pre>
    cos_values <- pmin(pmax(cos_values, -1), 1) # Clamp values to [-1, 1]
    theta_rest <- acos(cos_values)</pre>
  } else {
    theta_rest <- numeric(0) # No additional angles if p \le 2
  # Return the polar representation
  return(c(R, theta1, theta_rest))
}
# Test the function on a matrix y
set.seed(123)
y <- matrix(rnorm(1000 * 5), nrow = 1000, ncol = 5)
polar_representation <- t(apply(y, 1, polaroid))</pre>
# Extract R values and theta values
R values <- polar representation[, 1]</pre>
```

```
theta_values <- if (ncol(polar_representation) > 1) polar_representation[, -1] else NULL

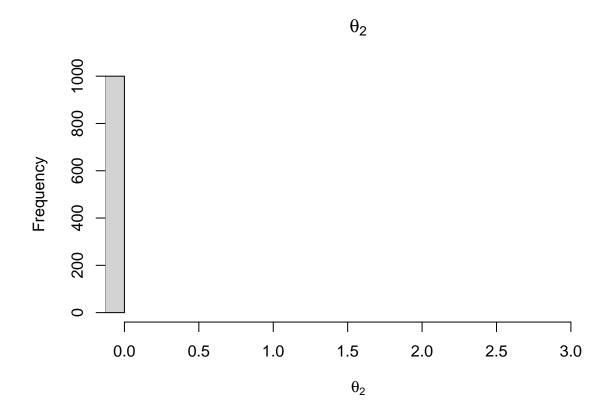
R_squared <- R_values^2
chisq_test <- ks.test(R_squared, "pchisq", df = 5)
chisq_test

##
## Asymptotic one-sample Kolmogorov-Smirnov test
##
## data: R_squared
## D = 0.020785, p-value = 0.7807
## alternative hypothesis: two-sided

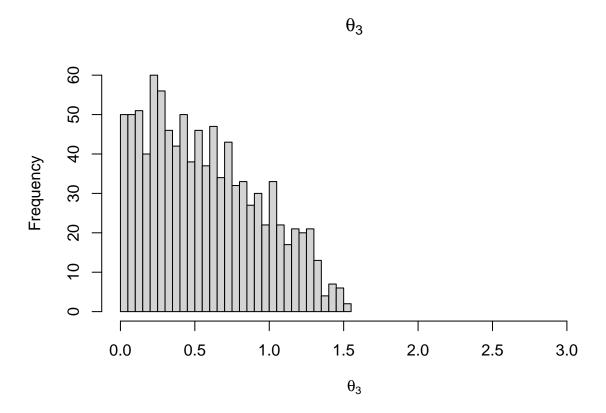
hist(theta_values[, 1], breaks = 30, main = expression(theta[1]), xlab = expression(theta[1]), xlim = c</pre>
```



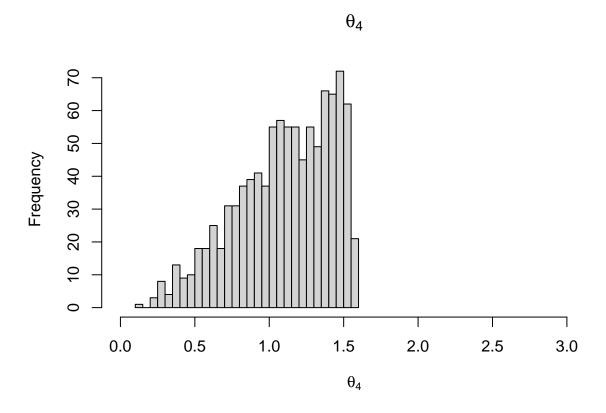
hist(theta\_values[, 2], breaks = 30, main = expression(theta[2]), xlab = expression(theta[2]), xlim = c



hist(theta\_values[, 3], breaks = 30, main = expression(theta[3]), xlab = expression(theta[3]), xlim = c



hist(theta\_values[, 4], breaks = 30, main = expression(theta[4]), xlab = expression(theta[4]), xlim = c



```
theta1_test <- ks.test(theta_values[, 1] / (2 * pi), "punif", min = 0, max = 1)
theta2_test <- ks.test(theta_values[, 2] / pi, "punif", min = 0, max = 1)</pre>
## Warning in ks.test.default(theta_values[, 2]/pi, "punif", min = 0, max = 1):
## ties should not be present for the one-sample Kolmogorov-Smirnov test
theta3_test <- ks.test(theta_values[, 3] / pi, "punif", min = 0, max = 1)</pre>
theta4_test <- ks.test(theta_values[, 4] / pi, "punif", min = 0, max = 1)</pre>
list(theta1_test = theta1_test, theta2_test = theta2_test,
     theta3_test = theta3_test, theta4_test = theta4_test)
## $theta1_test
##
    Asymptotic one-sample Kolmogorov-Smirnov test
##
##
## data: theta_values[, 1]/(2 * pi)
## D = 0.041297, p-value = 0.06602
## alternative hypothesis: two-sided
##
##
## $theta2_test
##
  Asymptotic one-sample Kolmogorov-Smirnov test
```

```
##
## data: theta_values[, 2]/pi
## D = 1, p-value < 2.2e-16
## alternative hypothesis: two-sided
##
## $theta3_test
##
## Asymptotic one-sample Kolmogorov-Smirnov test
##
## data: theta_values[, 3]/pi
## D = 0.5565, p-value < 2.2e-16
## alternative hypothesis: two-sided
##
##
## $theta4_test
##
   Asymptotic one-sample Kolmogorov-Smirnov test
##
## data: theta_values[, 4]/pi
## D = 0.50011, p-value < 2.2e-16
## alternative hypothesis: two-sided
```