

HW5

2024-09-29

Status

- Q1:
- Q2: DONE
- Q3: WIP
- Q4: WIP
- Q5: DONE
- Q6: DONE
- Q7: DONE

Homework 5

Due October 15

Q1

Q:

Use the `state.x77` data matrix and the `tapply()`, and separately, the `aggregate()` function to obtain

a.

The mean per capita income of the states in each of the four regions defined by the factor `state.region`,

b.

The maximum illiteracy rates for states in each of the nine divisions defined by the factor `state.division`,

c.

The number of states in each region,

d.

The names of the states in each division,

e.

The median high school graduation rates for groups of states defined by combinations of the factors state.region and state.size.

A:

a.

b.

c.

d.

e.

Q2

Q:

For a sample x_1, x_2, \dots, x_n , the MAD estimator of scale is defined as $1.4826 \cdot \text{median}\{|x_i - \bar{x}|\}$ where $\bar{x} = \text{median}\{x_i\}$

Use the matrix object `mtcars` to compute the MAD estimator of scale for the columns in `mtcars`

a.

Using the `apply()` function with the `mad()` function

b.

By calculating it directly from the definition (i.e., not using the `mad()` function). You may use the `apply()` and the `sweep()` functions but avoid using any loops

A:

a.

```
data(mtcars)

mad <- function(x) {
  median <- median(x, na.rm = TRUE)
  mad <- median(abs(x - median), na.rm = TRUE)
  1.4826 * mad
}

madDf <- apply(X = mtcars,
               MARGIN = 2,
               FUN = mad)

madDf
```

```
##      mpg      cyl      disp      hp      drat      wt
##  5.4114900  2.9652000 140.4763500  77.0952000  0.7042350  0.7672455
##      qsec      vs      am      gear      carb
##  1.4158830  0.0000000  0.0000000  1.4826000  1.4826000
```

b.

```
madDirect <- function(x) {
  median <- median(x, na.rm = TRUE)
  deviations <- abs(sweep(x = as.matrix(x),
                          MARGIN = 2,
                          STATS = median)
  )
  mad <- median(deviations, na.rm = TRUE)
  1.4826 * mad
}
```

```

}

madDdf <- apply(X = mtcars,
                MARGIN = 2,
                FUN = madDirect)

```

```
madDdf
```

```

##      mpg      cyl      disp      hp      drat      wt
##  5.4114900  2.9652000 140.4763500  77.0952000  0.7042350  0.7672455
##      qsec      vs      am      gear      carb
##  1.4158830  0.0000000  0.0000000  1.4826000  1.4826000

```

Q3

Q:

Let $h(x, n) = 1 + x + x^2 + \dots + x^n = \sum_{i=0}^n x^i$. Answer the following questions

a.

Write code to do the above in a for loop.

b.

Rewrite the code that you wrote to use a while loop.

c.

For each of the $x = 0.3, 1.01$ evaluate the performance for $n = 500, 5000$ in terms of time taken by the software. To do so, wrap the code around the R function `system.time()` with the same code as above inside the parentheses. Report the values returned in the output as per the user time field.

d.

Compare the above with results obtained avoiding loops.

A:

a.

```
x <- function(x, n) {  
  result <- 0 # Initialize the result  
  for (i in 0:n) { # Loop from 0 to n (inclusive)  
    result <- result + x^i # Add x^i to the result  
  }  
  return(result) # Return the computed sum  
}
```

b.

```
y <- function(x, n) {  
  result <- 0 # Initialize the result  
  i <- 0 # Initialize the counter  
  
  while (i <= n) { # Loop while i is less than or equal to n  
    result <- result + x^i # Add x^i to the result  
    i <- i + 1 # Increment the counter  
  }  
  
  return(result) # Return the computed sum  
}
```

c.

$x = 0.3, 1.01$ evaluate the performance for $n = 500, 5000$

```
x_value <- 0.3
n_value <- 500

system.time(expr =
"sum_result <- x(x_value, n_value)
sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
system.time(
"sum_result <- y(x_value, n_value)
sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
x_value <- 0.3
n_value <- 5000

system.time(expr =
"sum_result <- x(x_value, n_value)
sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
system.time(
"sum_result <- y(x_value, n_value)
sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
x_value <- 1.01
n_value <- 500

system.time(expr =
"sum_result <- x(x_value, n_value)
sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
system.time(
  "sum_result <- y(x_value, n_value)
  sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
x_value <- 1.01
n_value <- 5000
```

```
system.time(expr =
  "sum_result <- x(x_value, n_value)
  sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
system.time(
  "sum_result <- y(x_value, n_value)
  sum_result"
)
```

```
##      user  system elapsed
##         0         0         0
```

```
# Define the function using a for loop
h_for <- function(x, n) {
  result <- 1 # Initialize the sum
  for (i in 1:n) {
    result <- result + x^i
  }
  return(result)
}
```

```
# Define the function using a while loop
h_while <- function(x, n) {
  result <- 1 # Initialize the sum
  i <- 1 # Start from the first power
  while (i <= n) {
    result <- result + x^i
    i <- i + 1
  }
  return(result)
}
```

```
# Set values of x and n
x_values <- c(0.3, 1.01)
n_values <- c(500, 5000)
```

```
# Loop through each combination of x and n, measuring time
```

```

results <- data.frame()

for (x in x_values) {
  for (n in n_values) {
    # Measure time for the for loop
    time_for <- system.time(h_for(x, n))["user.self"]

    # Measure time for the while loop
    time_while <- system.time(h_while(x, n))["user.self"]

    # Store the results
    results <- rbind(results, data.frame(x = x, n = n, Time_for = time_for, Time_while = time_while))
  }
}

```

results

```

##           x      n Time_for Time_while
## user.self 0.30  500         0         0
## user.self1 0.30 5000         0         0
## user.self2 1.01  500         0         0
## user.self3 1.01 5000         0         0

```

d.

Q4

Q:

The Lotka-Volterra model for a predator-prey system assumes that $x(t)$ is the number of prey animals at the start of year t and that $y(t)$ is the number of predators at the start of year t . Then the number of prey animals and predators at the end of the following year is given by:

$$x(t+1) = x(t) + b_x x(t) - d_x x(t)y(t)$$

$$y(t+1) = y(t) + b_y d_x x(t)y(t) - d_y y(t)$$

Where b_x , b_y , d_x and d_y are as follows: - b_x is the natural birth rate of the prey animals in the absence of predation - d_x is the death rate of prey animal in an encounter with the predator. - d_y is the natural death rate of the predators in the absence of food (prey animals) - b_y is the efficiency of turning predated animals into predators

Let $b_x = 0.04$, $d_x = 0.0005$, $b_y = 0.1$, and $d_y = 0.2$.

Suppose that there were 4000 animal of the prey variety at the beginning of the time period. Also, suppose that there were only 100 predators. Write a while loop to show the predator-prey system as long as there are over 3900 prey animals. Save the prey/predator output in a list and plot them using lines on the same plot.

A:

```
# Set initial parameters
bx <- 0.04 # Natural birth rate of prey
dx <- 0.0005 # Death rate of prey in an encounter with predator
by <- 0.1 # Efficiency of turning predated animals into predators
dy <- 0.2 # Natural death rate of predators

# Initial conditions
prey <- 4000 # Initial number of prey
predator <- 100 # Initial number of predators

# Create lists to store prey and predator values over time
prey_list <- c(prey)
predator_list <- c(predator)

# While loop to simulate the predator-prey system
while (prey > 3900) {
  # Update prey and predator populations using Lotka-Volterra model
  new_prex <- prey + bx * prey - dx * prey * predator
  new_predator <- predator + by * dx * prey * predator - dy * predator

  # Append the new populations to the lists
  prey_list <- c(prey_list, new_prex)
  predator_list <- c(predator_list, new_predator)

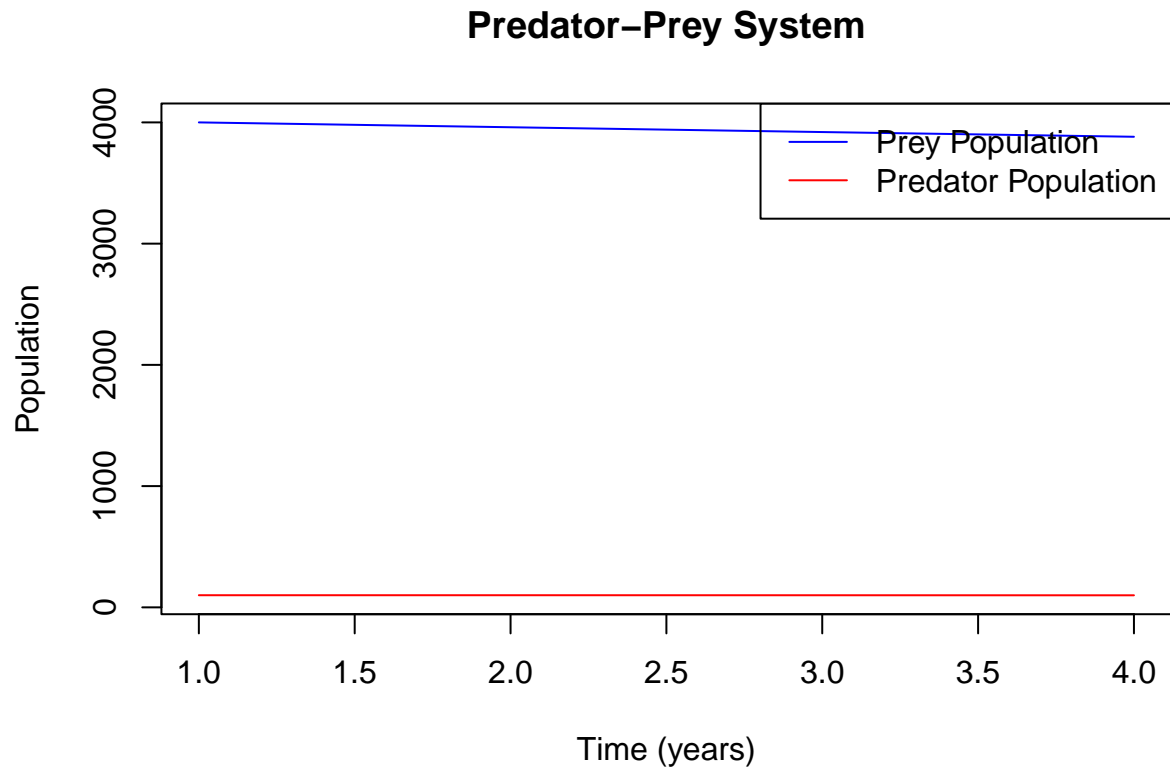
  # Update the current prey and predator populations for the next iteration
  prey <- new_prex
  predator <- new_predator
}
```

```

}

# Plot the results
plot(preym_list, type="l", col="blue", ylim=c(min(c(preym_list, predator_list)), max(c(preym_list, predator_list))),
      xlab="Time (years)", ylab="Population", main="Predator-Prey System")
lines(predator_list, col="red")
legend("topright", legend=c("Prey Population", "Predator Population"), col=c("blue", "red"), lty=1)

```



Q5

Q:

The game of craps is played as follows: first, Player 1 rolls two six-sided die; let x be the sum of the die on the first roll. If $x = 7$ or $x = 11$, then Player 1 wins, otherwise the player continues rolling until (s)he gets x again, in which case also Player 1 wins, or until (s)he gets 7 or 11, in which case (s)he loses. Write R code to simulate the game of craps. You can simulate the roll of a fair die using the `sample()` function in R.

A:



Figure 1: Saw what I did there?

```
ohCrap <- function() {  
  roll1 <- sum(sample(1:6, 2, replace = TRUE))  
  if (roll1 == 7 || roll1 == 11) {  
    return("Player 1 wins on the first roll!")  
  } else {  
    point <- roll1  
    repeat {  
      rollN <- sum(sample(1:6, 2, replace = TRUE))  
      if (rollN == point) {  
        return("Player 1 wins!")  
      }  
      if (rollN == 7 || rollN == 11) {  
        return("Player 1 loses!")  
      }  
    }  
  }  
}
```

```
}  
  
print("Let's play a game!")
```

```
## [1] "Let's play a game!"
```

```
rep(ohCrap(), 10)
```

```
## [1] "Player 1 loses!" "Player 1 loses!" "Player 1 loses!" "Player 1 loses!"  
## [5] "Player 1 loses!" "Player 1 loses!" "Player 1 loses!" "Player 1 loses!"  
## [9] "Player 1 loses!" "Player 1 loses!"
```

Orange you glad I went with the (jig)Saw joke instead of a poop joke?

Q6

Q:

Suppose that $(x(t), y(t))$ has polar coordinates given by $(\sqrt{t}, 2\pi t)$. Write code to plot the curve $(x(t), y(t))$ for $t \in [0, 1]$.

A:

Polar to Cartesian coordinates, r radius $x = r \cos \theta$ $y = r \sin \theta$

Where $r = \sqrt{t}$ Add $\theta = 2\pi t$

Utilizing this conversion, we then have:

$$x(t) = \sqrt{t} \cdot \cos(2\pi t) \quad y(t) = \sqrt{t} \cdot \sin(2\pi t)$$

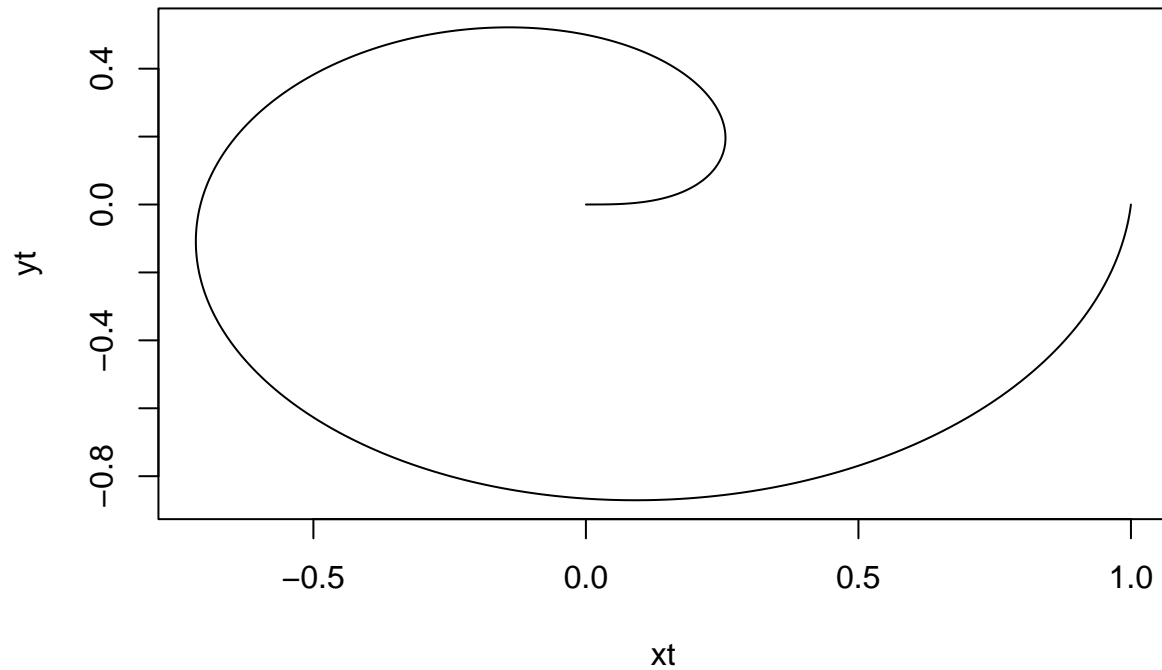
```
t <- seq(from = 0, to = 1, by = 0.001)

# xt <- sqrt(t)
# yt <- 2 * pi * t

xt <- sqrt(t) * cos(2 * pi * t)
yt <- sqrt(t) * sin(2 * pi * t)

plot(x = xt,
     y = yt,
     type = "l",
     main = "Look, A Spiral!")
```

Look, A Spiral!



Here's a for loop of it.

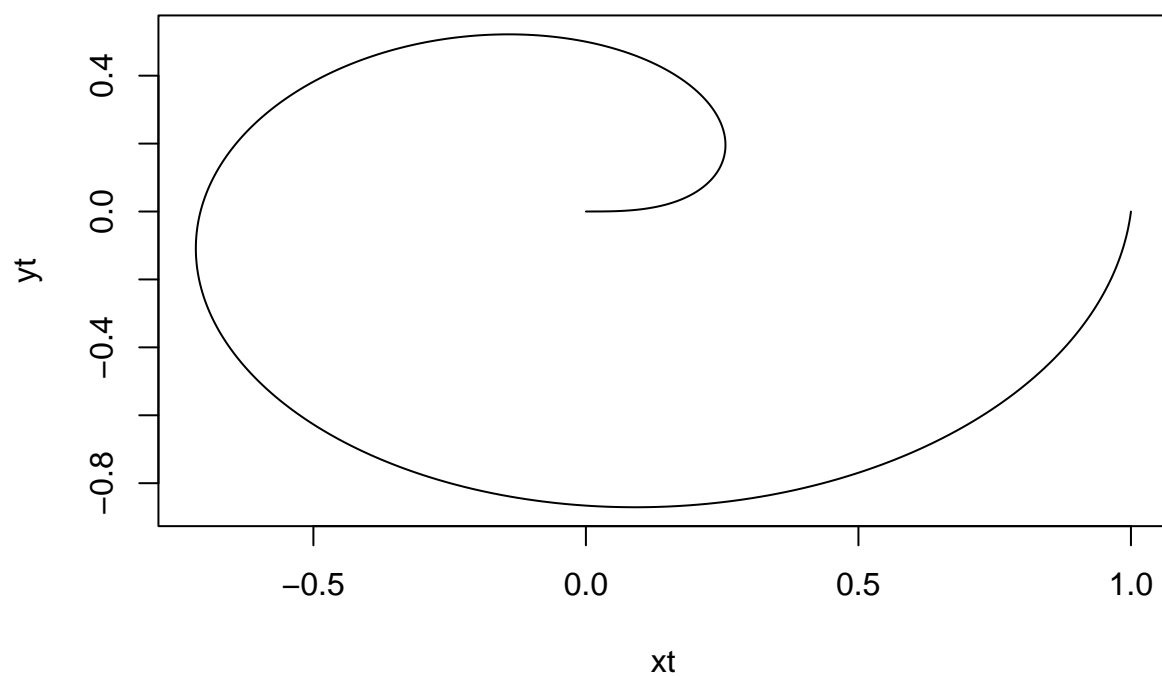
```
t <- seq(from = 0, to = 1, by = 0.001)

xt <- numeric(length(t))
yt <- numeric(length(t))

for (i in 1:length(t)) {
  xt[i] <- sqrt(t[i]) * cos(2 * pi * t[i])
  yt[i] <- sqrt(t[i]) * sin(2 * pi * t[i])
}

plot(x = xt,
     y = yt,
     type = "l",
     main = "Look, A Spiral, Again!")
```

Look, A Spiral, Again!



Q7

Q:

Consider the following code:

```
x <- matrix(rnorm(n = 500), ncol = 5)
varx <- var(x)
```

Starting with varx, use two applications of the sweep() function, one dividing each row of the matrix and the other dividing each column, of a covariance matrix to obtain R, the correlation matrix.

A:

```
# start
x <- matrix(rnorm(n = 500), ncol = 5)
varx <- var(x)

# the problem
std_devs <- sqrt(diag(varx))

rowSweep <- sweep(x = varx,
                  MARGIN = 1,
                  STATS = std_devs,
                  FUN = "/")

# 5x5 matrix
corR <- sweep(x = rowSweep,
             MARGIN = 2,
             STATS = std_devs,
             FUN = "/")

corR
```

```
##           [,1]           [,2]           [,3]           [,4]           [,5]
## [1,] 1.000000000 0.002398594 0.005244504 0.006703068 0.02608854
## [2,] 0.002398594 1.000000000 0.113193414 0.007724686 -0.01020370
## [3,] 0.005244504 0.113193414 1.000000000 0.034009075 -0.15449342
## [4,] 0.006703068 0.007724686 0.034009075 1.000000000 0.01346448
## [5,] 0.026088545 -0.010203704 -0.154493418 0.013464482 1.000000000
```