Tutorial 1 – Giovanni Filomeno - 12315325

1. Exercise 1

 a) Compute by hand the resulting sequence of pseudo-random numbers

Since the part (b) asks for the period of the sequence, I will evaluate by hand sequence of pseudo-random numbers until I have again the same number. The general formula is $x_{n+1} = 7x_n \pmod{31}$ with $x_0 = 19$.

```
• x_1 = 7 \times 19 \mod 31 = 133 \mod 31 = 9
```

•
$$x_2 = 7 \times 9 \mod 31 = 63 \mod 31 = 1$$

•
$$x_3 = 7 \times 1 \mod 31 = 7 \mod 31 = 7$$

•
$$x_4 = 7 \times 7 \mod 31 = 49 \mod 31 = 18$$

•
$$x_5 = 7 \times 18 \mod 31 = 126 \mod 31 = 2$$

•
$$x_6 = 7 \times 2 \mod 31 = 14 \mod 31 = 14$$

•
$$x_7 = 7 \times 14 \mod 31 = 98 \mod 31 = 5$$

•
$$x_8 = 7 \times 5 \mod 31 = 35 \mod 31 = 4$$

•
$$x_9 = 7 \times 4 \mod 31 = 28 \mod 31 = 28$$

•
$$x_{10} = 7 \times 28 \mod 31 = 196 \mod 31 = 10$$

•
$$x_{11} = 7 \times 10 \mod 31 = 70 \mod 31 = 8$$

•
$$x_{12} = 7 \times 8 \mod 31 = 56 \mod 31 = 25$$

•
$$x_{13} = 7 \times 25 \mod 31 = 175 \mod 31 = 20$$

•
$$x_{14} = 7 \times 20 \mod 31 = 140 \mod 31 = 16$$

•
$$x_{15} = 7 \times 16 \mod 31 = 112 \mod 31 = 19$$

b) What is the period of the sequence?

Since I obtain again 19 after 15 iterations, the period is 15.

Why does it not have maximum period m-1=30?

Because the period 31-1=30 is the theoretical maximum period given m=31, however not every multiplier a is a primitive root of mod 31. Finding a period of 15 leads to the following congruence $7^{15} \equiv 1 \pmod{31}$ which is true if I compute it by hand (4747561509943-1)/31=153147145482.

Find a value for a such that the generator has maximum period

I can use the formula $a^k \equiv 1 \pmod{31}$ where k=30 and I try some proper divisor of 30 for example $d=\{3,5,6,\ldots\}$. Doing this, I found that a=3 which leads to the following formula $x_{n+1}=3x_n \pmod{31}$.

2. Exercise 2

a) Implement the linear congruential generator

The screenshot of the R-code is provided in Figure 1.

```
ex2.R ×
                                                                             =\Box
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                                                         Run 🕪 🕞 Source 🗸 🗏
  1 # 1) Define the LCG function
  2 - lcg <- function(a, c, m, x0) {
      vals <- integer(0) # storing the generated value
      curr <- x0 # Initialization
  6 - repeat {
  7
        next_val <- (a * curr + c) %% m # Generating next value
  8
        vals <- c(vals, next_val) # Store the generated vyalue
  9
        curr <- next_val # Update the state
 10
 11 -
        if (curr == x0) { # Stop in case original value found
 12
           break
 13 ^
 14 ^ }
 15
 16
      return(vals)
 17 ^ }
 18
 19 # Exercise 2.b
 20 seq_{lcg} < -lcg(a = 7, c = 0, m = 31, x0 = 19)
 21 # Printing the sequence
 22 seq_lcg
 23
```

Figure 1: R code

b) Run the function using values from Exercise 1

The code in exercise 2.a gives the following sequence:

```
> # Printing the sequence
> seq_lcg
[1] 9 1 7 18 2 14 5 4 28 10 8 25 20 16 19
> |
```

Figure 2: Sequence

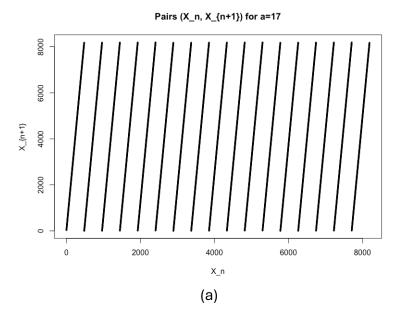
3. Exercise 3

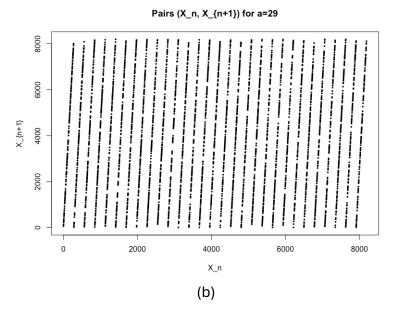
a) What is the period of each generated sequence?

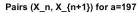
I used the same code of Exercise 2 three times, using the command *length()*. The results are: 8190 (A), 4095 (B), 4095 (C).

b) Plot all pairs of consecutive numbers and consider the autocorrelation for each of the three cases.

The Figure 3 shows the plot of all pairs of consecutive numbers for a=17, a=29, and a=197.







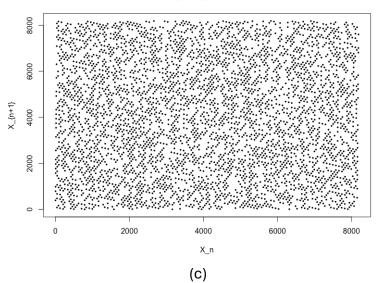


Figure 3: Plots of consecutive numbers

Additionally, as requested from the exercises, I also plot the auto-correlation, however, since the correlation in each sequence is pretty small, the three graphs look the same even though their periods differ (8190 vs 4095). Therefore, they do not add any meaningful information. However, just for sake of completeness, I also added in Figure 4 the plot coming from the function acf() for a=17.

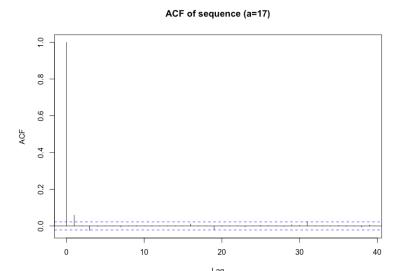


Figure 4: ACF plot for a=17

c) What is the best choice of multiplier a between (A), (B) and (C)?

I would take the multiplier of (A), a=17, since the period is larger, so it means a repetition comes later, even if visually it tends to form straight lines.

4. Exercise 4

a) Write function Fks(x)

```
Fks <- function(x, terms = 50) {
  if (x <= 0) return(0) # as requested in the exercise
  sum_val <- 0
  for (k in seq_len(terms)) { # Added a truncation as requested in the ex.
    sum_val <- sum_val + (-1)^(k-1) * exp(-2 * (k^2) * x^2)
  }
  # The series gives 1 - 2 * sumVal
  return(1 - 2 * sum_val)
}

# Vectorize as requested
Fks <- Vectorize(Fks)</pre>
```

Figure 5: Fks(x) function

b) Define function Kalpha(alpha)

```
Kalpha <- function(alpha) {
  if (alpha <= 0 || alpha >= 1) {
    stop("alpha must be in (0,1).")
  }
  target <- 1 - alpha  # we want Fks(K_alpha) = target
  f <- function(x) Fks(x) - target

# Bracking the root to avoid unstability
  sol <- uniroot(f, interval = c(1e-5, 5)) # using uniroot as suggested
  return(sol$root)
}</pre>
```

Figure 6: Kalpha(alpha) function

c) Write function Dno(A)

```
Dno <- function(A) {
    A_sorted <- sort(A)
    n <- length(A)

# D+ = max_{i} [i/n - X_(i)]
Dplus <- max( (seq_len(n))/n - A_sorted )
# D- = max_{i} [X_(i) - (i-1)/n]
Dminus <- max( A_sorted - (seq_len(n)-1)/n )

D <- max(Dplus, Dminus)
    return(D)
}</pre>
```

Figure 7: Dno(A) function

d) Write function *pval(A)*

```
pval <- function(A) {
    n <- length(A)
    d <- Dno(A)
    stat <- sqrt(n)*d
    return(1 - Fks(stat))
}</pre>
```

Figure 8: Dno(A) function

e) Write function ourkstest(A,alpha)

```
ourkstest <- function(A, alpha) {
    # Computing D using 4.c function
    D <- Dno(A)
    # Computing p-value using 4.d function
    p <- pval(A)
    # Making decision: p < alpha --> reject.
    decision <- if (p < alpha) "Reject H0" else "Do not reject H0"

result <- list(
    D = D, # Requested by ex.
    p.value = p, # Requested by ex.
    alpha = alpha, # not asked, but useful for the print
    decision = decision # not asked, but useful for reject/not reject test
)
    return(result)
}</pre>
```

Figure 9: ourkstest(A,alpha) function

f) Run the code

Figure 10: Run the two examples

```
ourkstest result, alpha=0.05:
ourkstest result, alpha=0.01:
                                       > print(res2)
> print(res1)
                                       [1] 0.2108352
                                                                             Compare with built-in ks.test:
[1] 0.2108352
                                                                              > print( ks.test(Asim, "punif") )
                                      $p.value
$p.value
                                                                                    Exact one-sample Kolmogorov-Smirnov
                                      [1] 0.02347071
[1] 0.02347071
                                                                             data: Asim
                                                                             D = 0.21084, p-value = 0.01986
                                       $alpha
$alpha
                                                                             alternative hypothesis: two-sided
                                       [1] 0.05
[1] 0.01
                                       $decision
$decision
                                       [1] "Reject H0"
[1] "Do not reject H0"
```

Figure 11: Output of the previous code

My functions give in both tests the same final decision w.r.t. to buil-in function. At $\alpha=0.01$ I do not reject H0 ($p\approx0.0235$ for my functions, $p\approx0.0199$ for the build-in. Both >0.01). At $\alpha=0.05$ I reject H0 ($p\approx0.0235$ for my functions, $p\approx0.0199$ for the build-in. Both <0.05). The numerical difference comes from the truncation (n=50) I have in the function.

5. Exercise 5

```
qcg <- function(d, a, c, m, x0) {
  vals <- integer(0) # Initialization
  curr <- x0 # First value

repeat {
    # Generate next value via quadratic recurrence
    next_val <- (d * curr^2 + a * curr + c) %% m

    # Store the value found
    vals <- c(vals, next_val)

# Update state
  curr <- next_val

# Stop if I find a period
  if (curr == x0) {
    break
  }
}

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

Figure 12: Quadratic Congruential Generator function

6. Exercise 6

a) Find d and a

To find the optimal parameters, I simply hard-coded a double for loop (see Figure 13), which gives me d=2 and a=3.

```
for (d_try in 1:100) { # loop over numbers
  for (a_try in seq(1,99,2)) { # loop over numbers
    p <- period_length(d_try, a_try)
    if (p > best_period) {
       best_period <- p
       best_params <- c(d_try, a_try)

    if (best_period == 65536) { # Add a breaking
       break
    }
  }
}
if (best_period == 65536) {
  break
}
</pre>
```

Figure 13: Double for to find the parameters

b) Perform a test

After finding the requested a and d, I generated the sequence, and I normalized the sequence over m and then I applied the build-in Kolmogorov Smirnov test which gives me D=1.5259e-05 and p-value=1.

c) Plot 1000 pairs

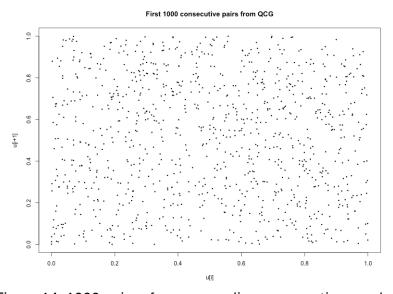


Figure 14: 1000 pairs of corresponding consecutive numbers