

# Sheet03

May 9, 2022

## 1 Exercise 5

### 1.1 b)

```
[1]: import numpy as np
from project_a1.random import LCG
import matplotlib.pyplot as plt

# Task b)

def check_array (value, array):
    for i in range(len(array)):
        if array[i] == value:
            return False
    return True

def periodtest(a1):
    seed1 = 0
    i = 0
    gen = LCG(seed = seed1, a = a1, c = 3, m = 1024)
    x = np.zeros(1024)
    gen.advance()
    while i <= 1024 and (gen.state != seed1) and check_array(gen.state, x):
        x[i] = gen.state
        gen.advance()
        i = i + 1
    return i

print("a = 10: periodlength = ", periodtest(10))
print("a = 1: periodlength = ", periodtest(1))
print("a = 69: periodlength = ", periodtest(69))
print("a = 5: periodlength = ", periodtest(5))
print("a = 3: periodlength = ", periodtest(3))
print("a = 4: periodlength = ", periodtest(4))
```

```
a = 10: periodlength = 10
a = 1: periodlength = 1023
a = 69: periodlength = 1023
```

```

a = 5: periodlength = 1023
a = 3: periodlength = 511
a = 4: periodlength = 5

```

Diffrent parameters  $a$  result in diffrent period lenghts. The Maximum period length is equal to the parameter  $m = 1024$  and is reached for  $a$ 's that fulfill the following conditions: -  $a - 1$  and  $m$  are divisible by 4 - each prime factor of  $m$  divides  $a - 1$  The prime factor of  $m = 1024$  is 2  $\rightarrow a - 1$  must be divisible by 2 and 4. E.g.  $a = 69$  an  $a = 5$  are such ideal numbers.

## 1.2 d) - f)

```

[2]: # Tasks d) - f)

# Numpy generator
rng = np.random.default_rng(420)
x, y, z = rng.uniform(size = (3, 10000))

# Our LCG
gen = LCG(seed = 0, a = 1601, c = 3456, m = 10000)           # Exercise
    ↪ Parameters
x2, y2, z2 = gen.uniform(size = (3, 10000))

good_gen = LCG(seed = 0, a = 625, c = 6571, m = 31104)      # Good Parameters
x3, y3, z3 = good_gen.uniform(size = (3, 10000))

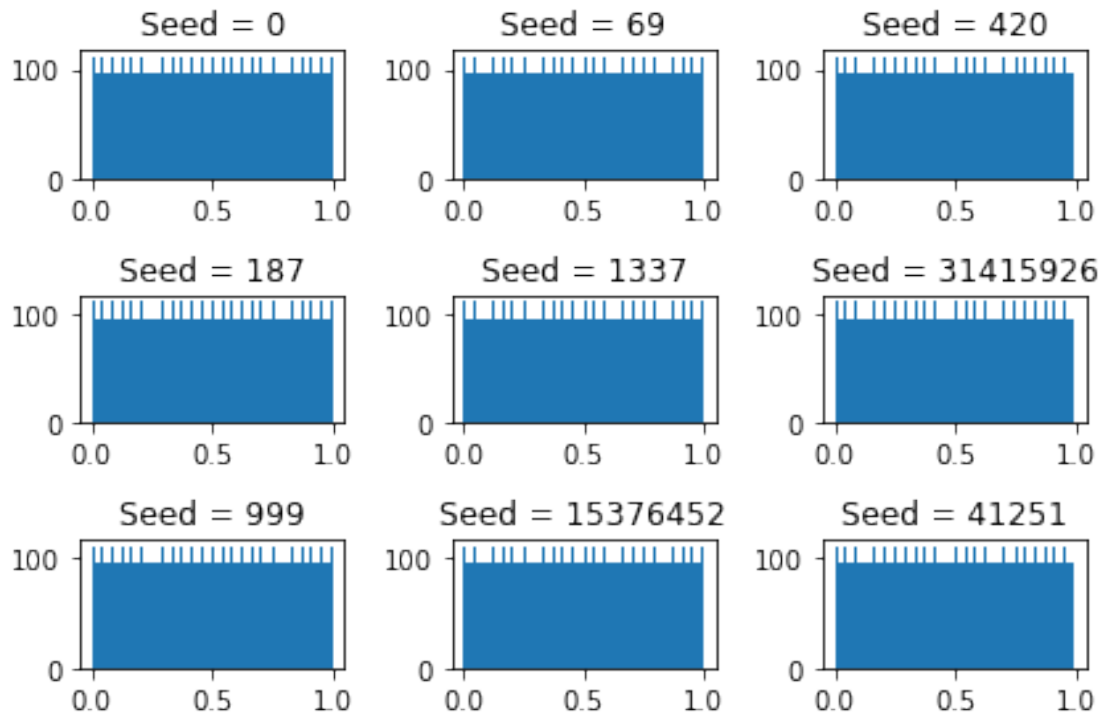
# Dependency on start value

values = [0, 69, 420, 187, 1337, 31415926, 999, 15376452, 41251]
fig0 = plt.figure()

for i in range(len(values)):
    ax = fig0.add_subplot(3, 3, i+1)
    tempgen = LCG(seed = values[i], a = 1601, c = 3456, m = 10000)
    ax.hist(tempgen.uniform(size = 10000), bins = 100)
    ax.set_title(f"Seed = {values[i]}")

fig0.tight_layout()
plt.show()

```



As it can be seen in the graphic different start values (seeds) do not lead to a different behaviour of the generator.

In the following the given LCG is compared to *numpy*'s generator and the same LCG with better parameters.

```
[3]: # Comparison of our LCG vs numpy.random

fig = plt.figure()

ax1 = fig.add_subplot(2, 3, 1)
ax1.hist(y2, bins = 100)           # Histogram of our "random" numbers
ax1.set_title("LCG")

ax2 = fig.add_subplot(2, 3, 4)
ax2.scatter(
    x2, y2,
    s=5,
    #smaller points
    alpha=0.3,
    # 70% transparency
)

ax3 = fig.add_subplot(2, 3, 2)
```

```

ax3.hist(y, bins = 100)                # histogram of numpy's random numbers
ax3.set_title("numpy.random")

ax4 = fig.add_subplot(2, 3, 5)
ax4.scatter(
    x, y,
    s=5,
    #smaller points
    alpha=0.3,
    #70% transparency
)

ax5 = fig.add_subplot(2, 3, 3)
ax5.hist(y3, bins = 100)                # histogram of numpy's random numbers
ax5.set_title("LCG w/ good params")

ax6 = fig.add_subplot(2, 3, 6)
ax6.scatter(
    x3, y3,
    s=5,
    #smaller points
    alpha=0.3,
    #70% transparency
)

fig.tight_layout()

# 3D Scatter-Plots

plt.show()

fig2 = plt.figure()

ax = fig2.add_subplot(1, 3, 1, projection='3d')

ax.scatter(
    x2, y2, z2,
    s=5,
    # smaller points
    alpha=0.3,
    # 70% transparency
)
ax.set_title("LCG")
# set the orientation of the 3d axis

ax.view_init(elev=30, azim=20)

```

```

ax2 = fig2.add_subplot(1, 3, 2, projection='3d')

ax2.scatter(
    x, y, z,
    s=5,
    # smaller points
    alpha=0.3,
    # 70% transparency
)
ax2.set_title("numpy.random")

# set the orientation of the 3d axis

ax2.view_init(elev=30, azimuth=20)

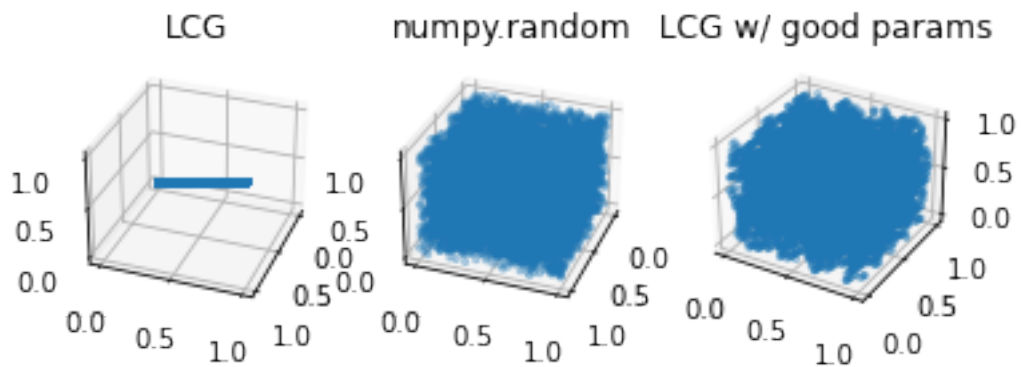
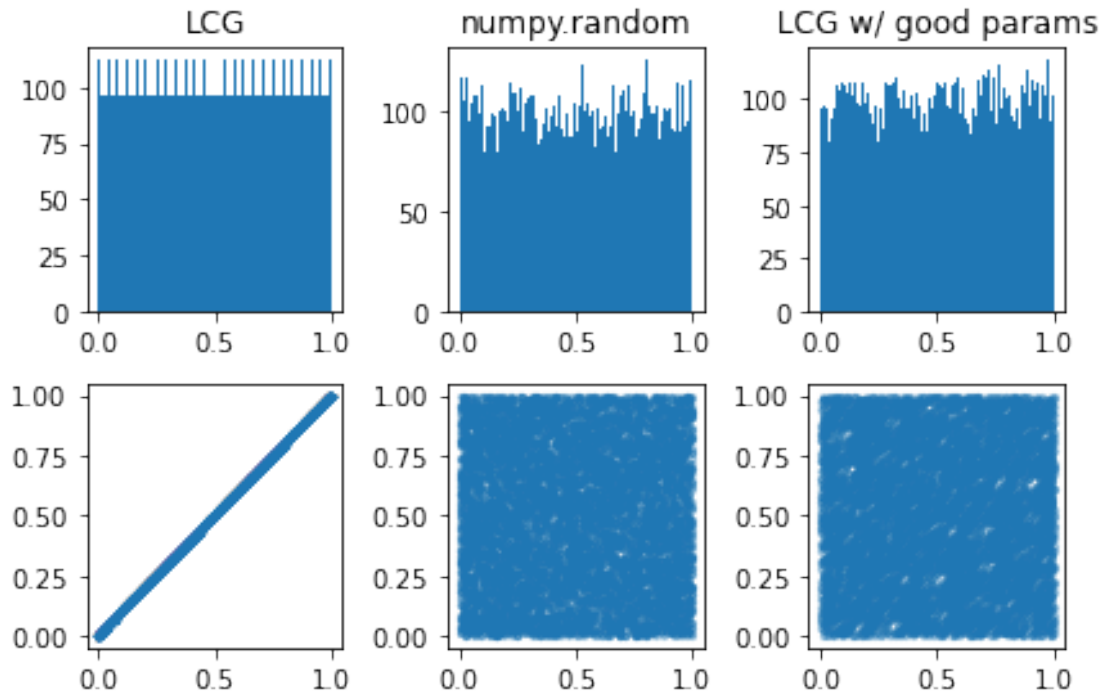
ax3 = fig2.add_subplot(1, 3, 3, projection='3d')

ax3.scatter(
    x3, y3, z3,
    s=5,
    # smaller points
    alpha=0.3,
    # 70% transparency
)
ax3.set_title("LCG w/ good params")
# set the orientation of the 3d axis

ax.view_init(elev=30, azimuth=20)

plt.show()

```



The plots on the left side show, that the given LCG ( $a = 1601$ ,  $c = 3456$  and  $m = 10000$ ) does not match the requirements of a good random generator. The periodlength is too short for high amounts of random numbers.

## 2 Exercise 6

### 2.1 a)

$$f(x) = \begin{cases} Ne^{-x/\tau} & 0 \leq x < \infty \\ 0 & else \end{cases}$$

**Determination of normilization constant  $N$ :**  $\int_0^\infty N e^{-x/\tau} dx \stackrel{!}{=} 1$

$$\Leftrightarrow N[-\tau \cdot e^{-x/\tau}]_0^\infty = N \cdot \tau = 1$$

$$\Leftrightarrow N = \frac{1}{\tau}$$

**Transformation of uniform Distribution  $u$  to  $x$  with propability density  $f(x)$ :**  $u =$

$$F(x) = [-e^{-y/\tau}]_0^x = -e^{-x/\tau} + 1$$

$$\Leftrightarrow 1 - u = e^{-x/\tau} \quad |\ln(\dots)$$

$$1 - u \in [0, 1], \text{ because } u \in [0, 1]$$

$$\Leftrightarrow \ln(1 - u) = -x/\tau$$

$$\Leftrightarrow x = -\ln(1 - u) \cdot \tau$$

## 2.2 b)

$$f(x) = \begin{cases} Nx^{-n} & x_{\min} \leq x \leq x_{\max} \\ 0 & else \end{cases}$$

**Determination of normilization constant  $N$ :**  $\int_{x_{\min}}^{x_{\max}} Nx^{-n} dx \stackrel{!}{=} 1$

$$\Leftrightarrow N \left[ \frac{1}{1-n} x^{1-n} \right]_{x_{\min}}^{x_{\max}} = 1$$

$$\Leftrightarrow N \frac{1}{1-n} (x_{\max}^{1-n} - x_{\min}^{1-n}) = 1$$

$$\Leftrightarrow N = \frac{1-n}{x_{\max}^{1-n} - x_{\min}^{1-n}}$$

**Transformation of uniform Distribution  $u$  to  $x$  with propability density  $f(x)$ :**  $u =$

$$F(x) = N \left[ \frac{y^{1-n}}{1-n} \right]_{x_{\min}}^x$$

$$\Leftrightarrow \frac{u(1-n)}{N} = x^{1-n} - x_{\min}^{1-n}$$

$$\Leftrightarrow x^{1-n} = \frac{u(1-n)}{N} + x_{\min}^{1-n}$$

$$\Leftrightarrow x = \sqrt[1-n]{\frac{u(1-n)}{N} + x_{\min}^{1-n}} \quad | N = \frac{1-n}{x_{\max}^{1-n} - x_{\min}^{1-n}}$$

$$\Leftrightarrow x = \sqrt[1-n]{u(x_{\max}^{1-n} - x_{\min}^{1-n}) + x_{\min}^{1-n}}$$

$$\Leftrightarrow x = x_{\min} \sqrt[1-n]{u \left( \left( \frac{x_{\max}}{x_{\min}} \right)^{1-n} - 1 \right) + 1}$$

## 2.3 c)

$$f(x) = \frac{1}{1+x^2}$$

Note: pi does not show in the pdf for some reason, so weird blank spaces in pdf should be pi.

**Transformation of uniform Distribution  $u$  to  $x$  with propability density  $f(x)$ :**  $u =$   
 $F(x) = \frac{1}{2} \int_{-\infty}^x \frac{1}{1+y^2} dy$

$$\Leftrightarrow u = \arctan(x) + \frac{\pi}{2}$$

$$\Leftrightarrow x = \tan \left( u - \frac{\pi}{2} \right)$$

### 2.3.1 Interpretation of results in ‘distributions.pdf’ and ‘lcg.pdf’

Since the histogram matches the distributions curve, the methods we implemented seem to work fine :)