

$$\begin{array}{r|l|l}
 5 & 6 & 2 \\
 \hline
 3,5 & 5 & 8,5 \\
 \hline
 5 & 5 & 10
 \end{array}$$

Sheet03

May 10, 2022

Programming Exercises:

The modified project code can be found in the branch: 'imsodone' in the project git lab.

Exercise 5

a)

```
[1]: def advance(self):
      '''
      advance the state of this random generator by one step

      Blatt 3, Aufgabe 1
      '''

      # Fügen Sie hier Ihren Code ein, um den LCG korrekt zu implementieren

      self.state = (self.a * self.state + self.c) % self.m
```

b)

```
[2]: from project_c3 import random
import numpy as np
import matplotlib.pyplot as plt

a_array = np.linspace(0, 1000, 100, dtype = int)
c = 3
m = 1024
amount_random_numbers = 10000

random_numbers = [random.LCG(a, c=c, m=m).random_raw(size =
    ↳ amount_random_numbers) for a in a_array]

def max_len(arr):

    mp = {}
```

you miss important values for a here, use arange(1, 1024, 1)

```

maxDict = 0

#creates a dictionary where the key is the element and the value is the
→ index of the element
#if a double element occurs and its distance to the previous occurrence is
→ the biggest yet,
#we update the max distance

for i in range(len(arr)):

    if arr[i] not in mp.keys():
        mp[arr[i]] = i

    else:
        maxDict = max(maxDict, i-mp[arr[i]])

return maxDict

max_lengths = [max_len(random_numbers[s]) for s in range(len(random_numbers))]

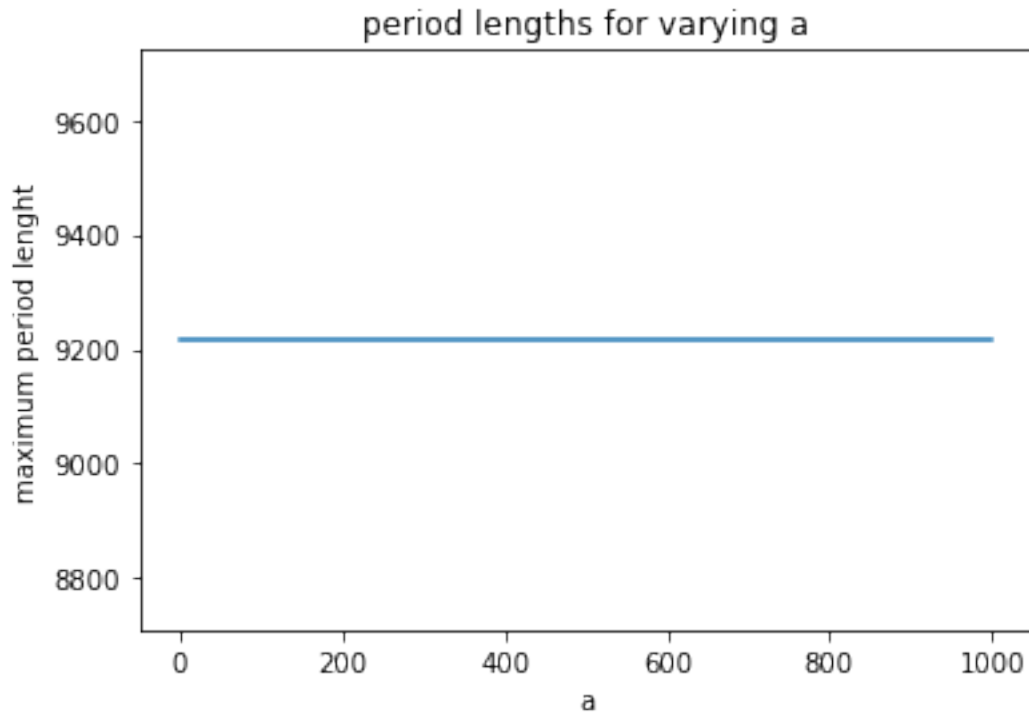
plt.figure()
plt.plot(a_array, max_lengths)
plt.xlabel('a')
plt.ylabel('maximum period length')
plt.title('period lengths for varying a')

None

```

you have to find
a win here.
As soon as a random
number repeats itself
you have found the period-
length.

See lecture for max. period length.



In this specific example, a does not have an influence on the period length \neq

c)

```
[3]: def uniform(self, low=0, high=1, size=None):
      """
      Draw uniform random numbers by converting them from the raw numbers'''
      raw = self.random_raw(size=size)

      # Fügen Sie hier ihren Code ein, um die rohen Zufallszahlen
      # zu kontinuierlich gleichverteilten Zufallszahlen zwischen
      # low und high zu transformieren

      u = np.divide(raw, self.m)

      result = u * (high - low) + low
      return result
```

d)

```
[4]: a = 1601
      c = 3456
      m = 10000
```

```

LCG_instance = random.LCG(a,c,m)

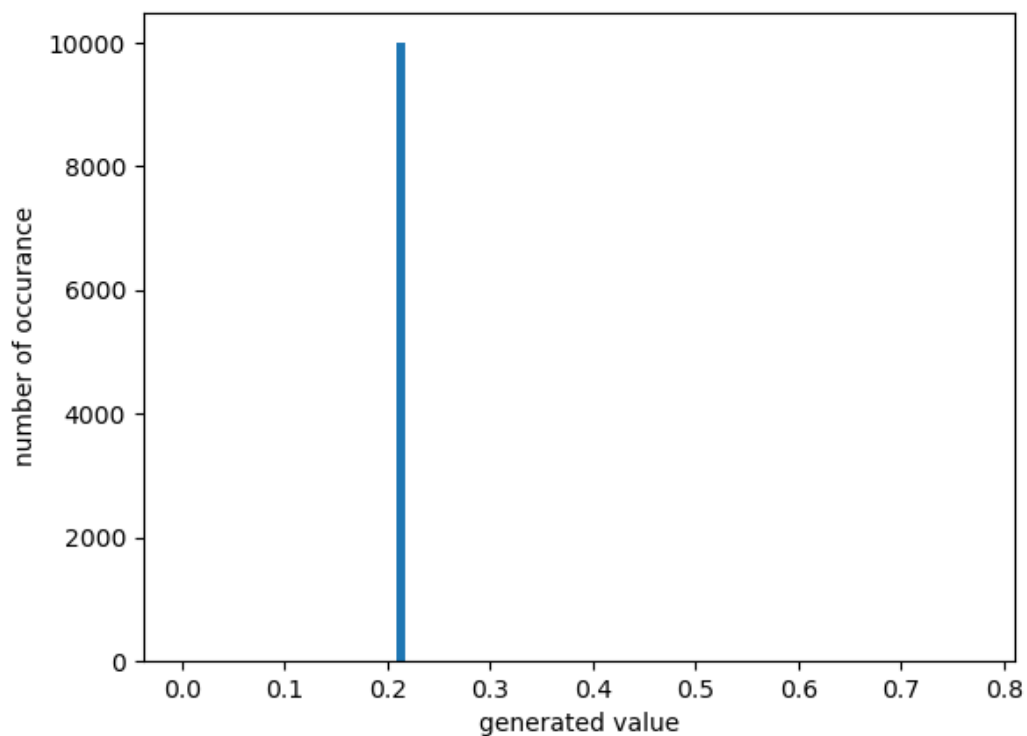
random_number_uniform = LCG_instance.uniform(low = 0, high = 1, size = 10000)

%matplotlib widget

plt.figure()
plt.hist(random_number_uniform, bins =100)
plt.xlabel('generated value')
plt.ylabel('number of occurance')

None

```



It does not meet the requirements, as the number 0.21562952 seems to generated unproportionally often compared to the other numbers.

```

[5]: seed = np.array([1, 2, 10, 100])

plt.figure()
for s in seed:
    LCG_instance = random.LCG(seed = s, a = a,c = c,m= m)

```

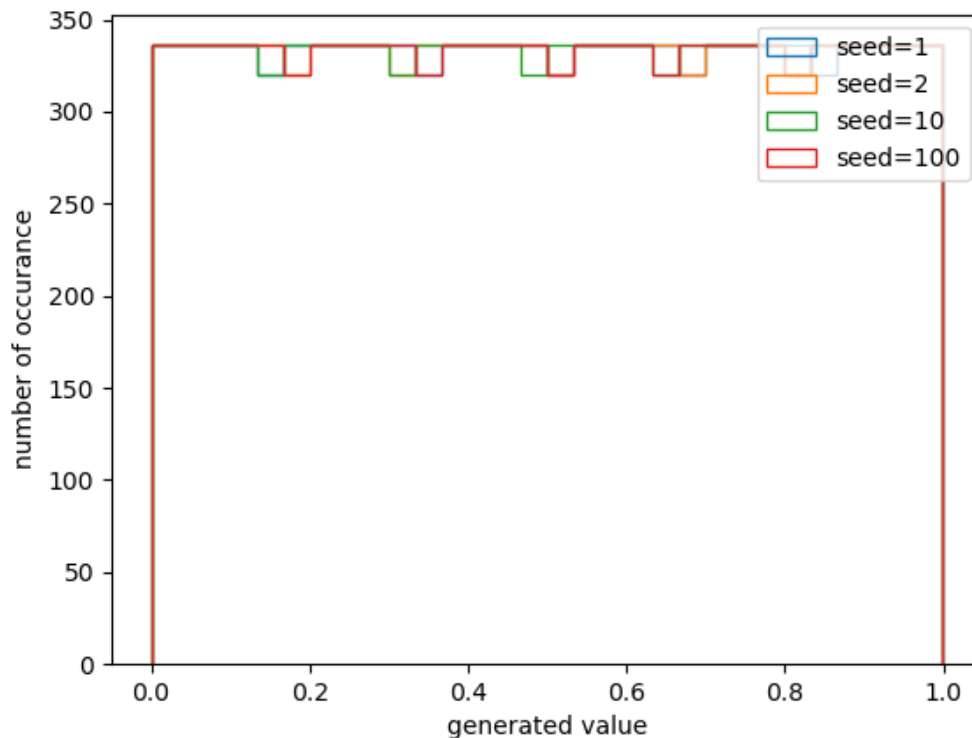
```

random_number_uniform = LCG_instance.uniform(low = 0, high = 1, size = 10000)

plt.hist(random_number_uniform, bins = 30, label = 'seed=' + str(s), histtype = 'step')
plt.xlabel('generated value')
plt.ylabel('number of occurrence')
plt.legend()

```

None



The seed value has a big influence on the maximum period length, with different seeds the distributions become almost uniform. However one can still see repeating patterns. (Drops at regular intervals)

c)

The lcg pdf shows that the implemented methods work (all Tests True), however one of the limits becomes apparent: As the lcg cannot generate negative numbers, the produced distributions will always have $x > 0$.

This is due to the small mistake in uniform().

e)

```
[6]: LCG_instance = random.LCG(a = a, c = c, m = m)

random_number_uniform = LCG_instance.uniform(low = 0, high = 1, size = 10000)

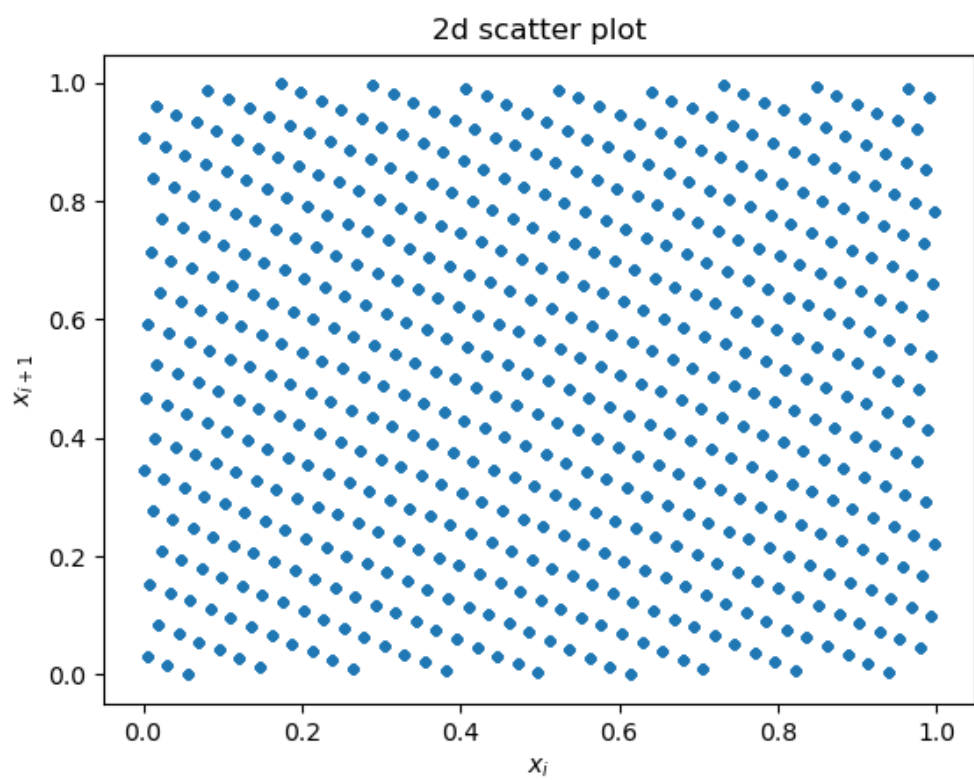
plt.figure()
plt.plot(random_number_uniform[:-1], random_number_uniform[1:], linestyle = 'None', marker='.')
plt.xlabel(r'$x_i$')
plt.ylabel(r'$x_{i+1}$')
plt.title('2d scatter plot')

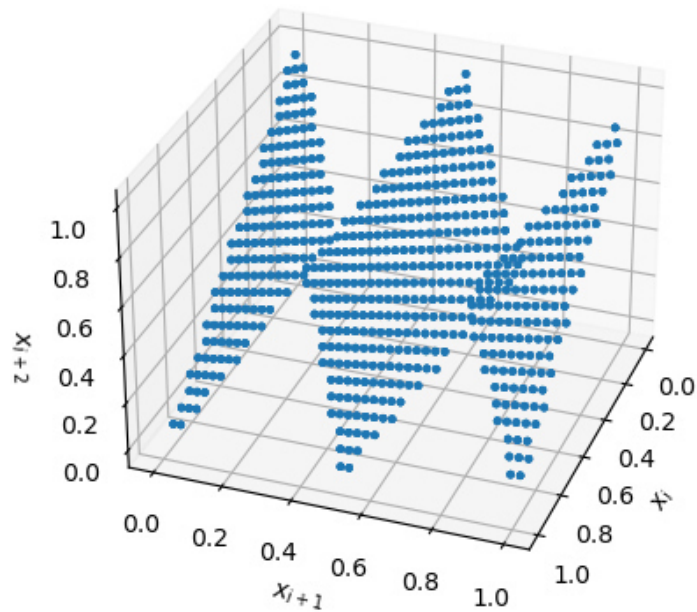
fig = plt.figure()
ax = fig.add_subplot(1, 1, 1, projection = '3d')

ax.scatter(
    random_number_uniform[:-2], random_number_uniform[1:-1],
    random_number_uniform[2:],
    s=5,
    alpha=0.3,
)

ax.view_init(elev=30, azimuth=20)
ax.set_xlabel(r'$x_i$')
ax.set_ylabel(r'$x_{i+1}$')
ax.set_zlabel(r'$x_{i+2}$')
```

None





The result does not meet the requirements. In the 2D Plot you can clearly see many parallel lines, meaning many value-pairs occur

The 3d plots shows a similar result, as we can clearly see parallel planes, meaning also many triplets occur

f)

```
[7]: rng = np.random.default_rng(0)

random_number_uniform = rng.random(10000)

plt.figure()
plt.plot(random_number_uniform[:-1], random_number_uniform[1:], linestyle = 'None', marker='.')
plt.xlabel(r'$x_i$')
plt.ylabel(r'$x_{i+1}$')
plt.title('2d scatter plot')

fig = plt.figure()
ax = fig.add_subplot(1, 1, 1, projection = '3d')
```



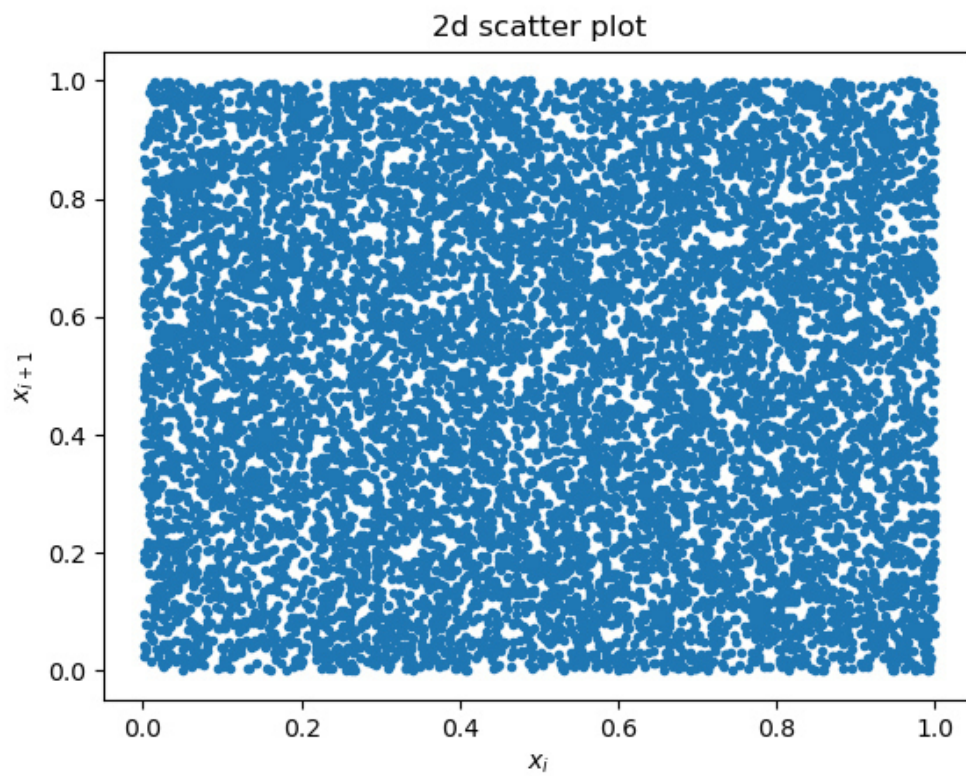
```

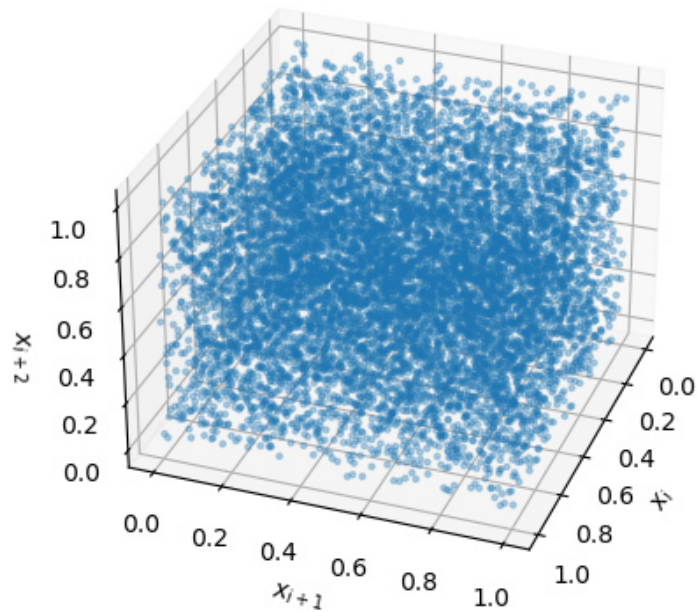
ax.scatter(
    random_number_uniform[:-2], random_number_uniform[1:-1],
    random_number_uniform[2:],
    s=5,
    alpha=0.3,
)

ax.view_init(elev=30, azimuth=20)
ax.set_xlabel(r'$x_i$')
ax.set_ylabel(r'$x_{i+1}$')
ax.set_zlabel(r'$x_{i+2}$')

```

None





The 2d and 3d plots show no apparent patterns and thus the generator meets the requirements for a good random number generator

✓ 3,5/5

Exercise 6:

Interpretation of the overview pdf:

The implemented distributions seem to work, as all the tests show true. Also the fitted functions match the underlying distribution, hinting at a correct implementation. The plots of the exponential and power distribution appear linear due to the chosen scale for the axis.

Code for Exercise 6:

```
[10]: def exponential(self, tau, size=None):
        """
        Draw exponentially distributed random numbers.

        Blatt 3, Aufgabe 2a)
        """
        # So können Sie ein array mit shape=size
        # mit standard gleichverteilten Zufallszahlen erzeugen
        u = self.uniform(size=size)
```

```

        # Fügen Sie hier den Code ein um Zufallszahlen aus der
        # angegebenen Verteilung zu erzeugen

        # dummy, so the code works. Can be removed / replaced
        values = -tau*np.log(1-u) ✓

    return values

def power(self, n, x_min, x_max, size=None):
    """
    Draw random numbers from a power law distribution
    with index n between x_min and x_max

    Blatt 3, Aufgabe 2b)
    """
    # Fügen Sie hier den Code ein um Zufallszahlen aus der
    # angegebenen Verteilung zu erzeugen

    # dummy, so the code works. Can be removed / replaced

    u = self.uniform(size=size)

    values = np.power(u*(x_max**(-n+1)-x_min**(-n+1))+x_min**(-n+1),1/(-n+1)) ✓

    return values

def cauchy(self, size=None):
    """
    Draw random numbers from a power law distribution
    with index n between x_min and x_max

    Blatt 3, Aufgabe 2c)
    """
    # Fügen Sie hier den Code ein um Zufallszahlen aus der
    # angegebenen Verteilung zu erzeugen

    # dummy, so the code works. Can be removed / replaced
    u = self.uniform(size=size)

    values = np.tan(np.pi*u-np.pi/2) ✓

    return values

```

Aufgabe 6

Die invertierten Funktionen werden dabei alle so auch im gitlab-Repo implementiert.

Exponentialverteilung

Die Exponentialfunktion lässt sich normieren auf:

$$\begin{aligned}\int_0^{\infty} N e^{-\frac{x}{\tau}} dx &\stackrel{!}{=} 1 \\ &= N \left[-\tau e^{-\frac{x}{\tau}} \right]_0^{\infty} \\ &= \tau\end{aligned}$$

$$\Rightarrow N = \tau^{-1}$$

Durch die Integration und Invertierung erhalten wir:

$$\int_0^{x'} \tau^{-1} e^{-\frac{x'}{\tau}} dx' = \left[-e^{-\frac{x'}{\tau}} \right]_0^{x'} = -e^{-\frac{x'}{\tau}} + 1$$

$$\Rightarrow x(u) = -\tau \ln(1 - u)$$

Potenzverteilung

Wie zuvor lässt sich die Funktion normieren zu:

$$\begin{aligned}\int_{x_{\min}}^{x_{\max}} N x^{-n} dx &\stackrel{!}{=} 1 \\ &= N \left[\frac{1}{1-n} \cdot x^{-n+1} \right]_{x_{\min}}^{x_{\max}} \\ &= N \frac{x_{\max}^{1-n} - x_{\min}^{1-n}}{1-n}\end{aligned}$$

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

und ebenfalls durch Integration und Invertierung erhalten wir:

$$\int_{x_{\min}}^{x'} N x'^{-n} dx' = \frac{1}{(x_{\max}^{1-n} - x_{\min}^{1-n})} (x^{-n+1} - x_{\min}^{-n+1})$$

$$\Rightarrow x(u) = \left(u \cdot (x_{\max}^{1-n} - x_{\min}^{1-n}) + x_{\min}^{-n+1} \right)^{\frac{1}{1-n}}$$

Cauchy-Verteilung

Die Cauchy-Verteilung muss nicht normiert werden, daher erhalten wir:

$$\int_{-\infty}^x \frac{1}{\pi(1+x'^2)} dx' = \frac{1}{\pi} \left(\arctan x + \frac{\pi}{2} \right)$$

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

In []:



~~8/5~~, nice!
you can consider
presenting this

Information:

Exercise: Linear-Kongruent

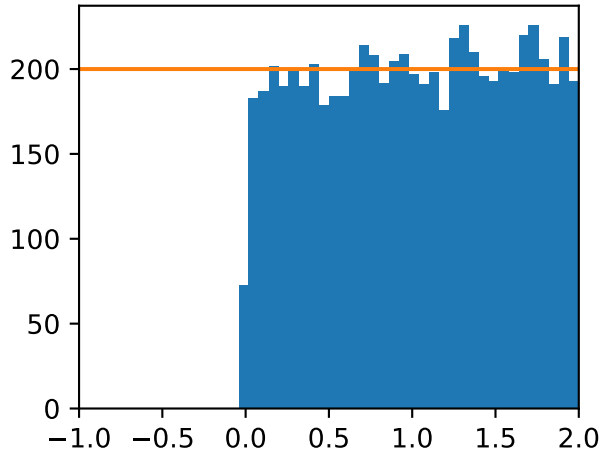
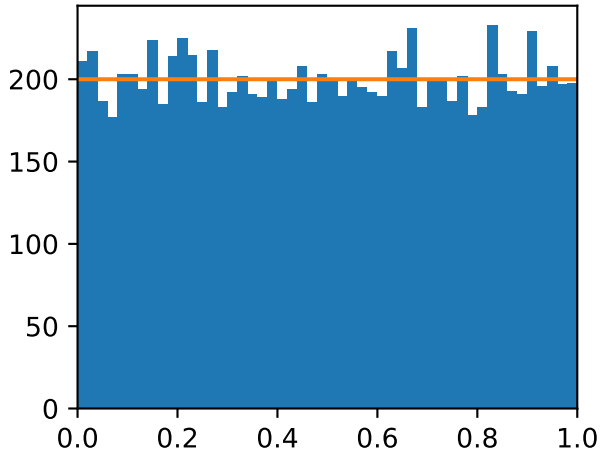
Group name: project_c3

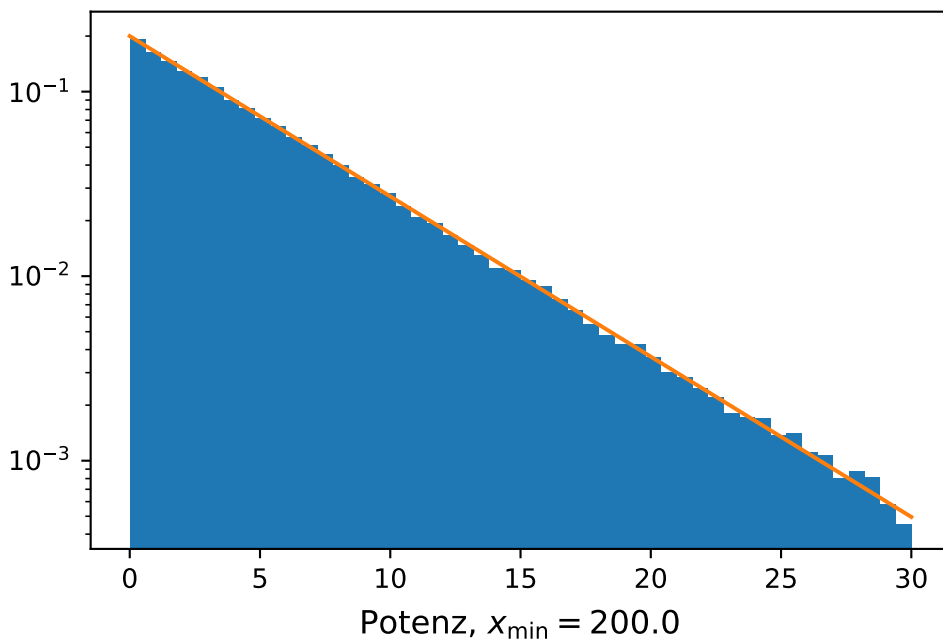
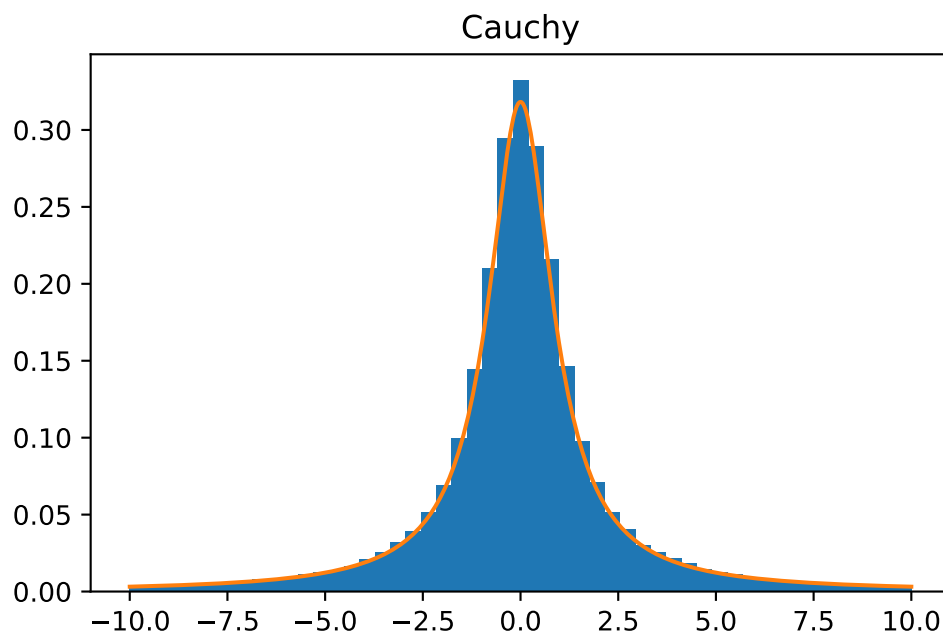
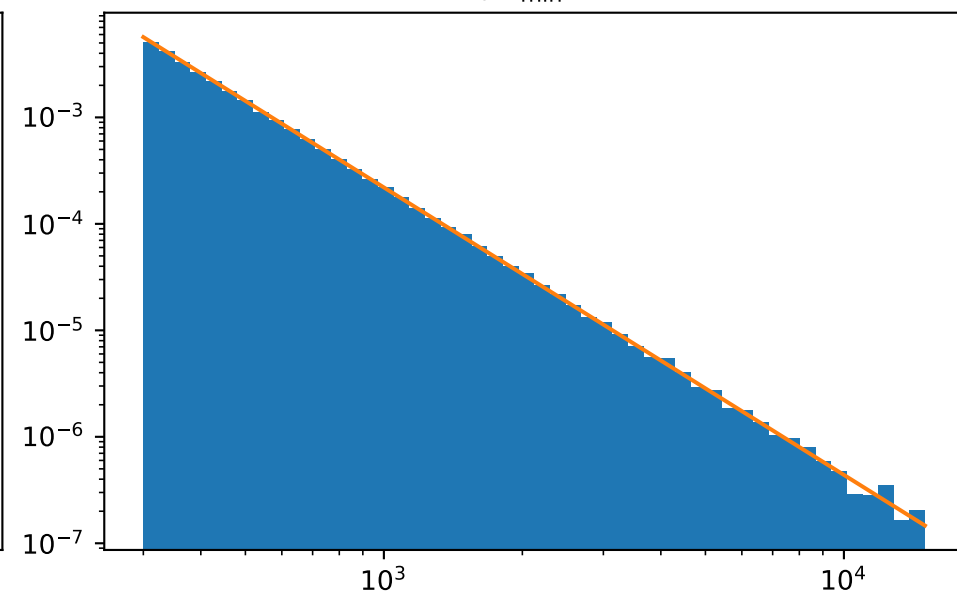
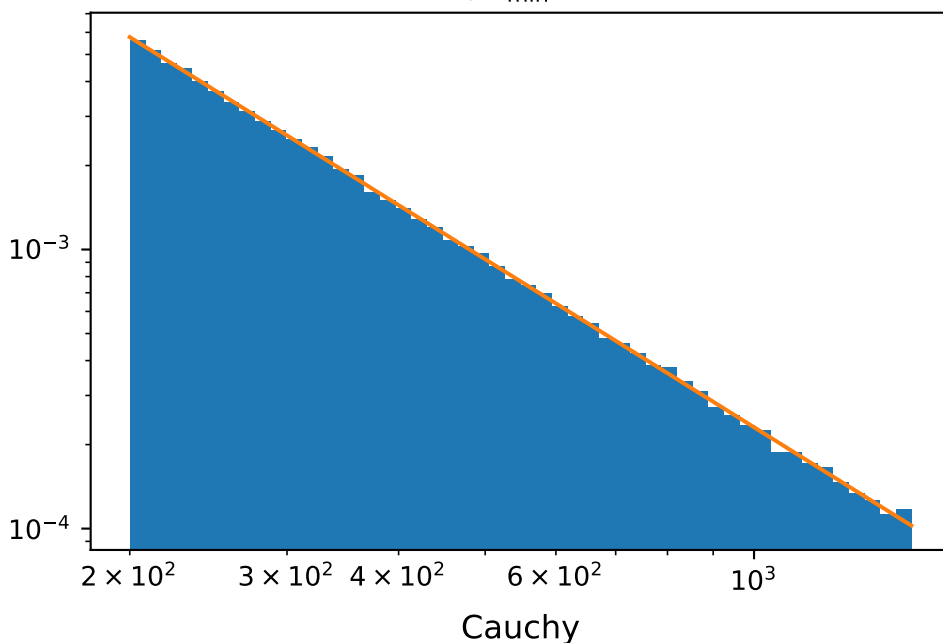
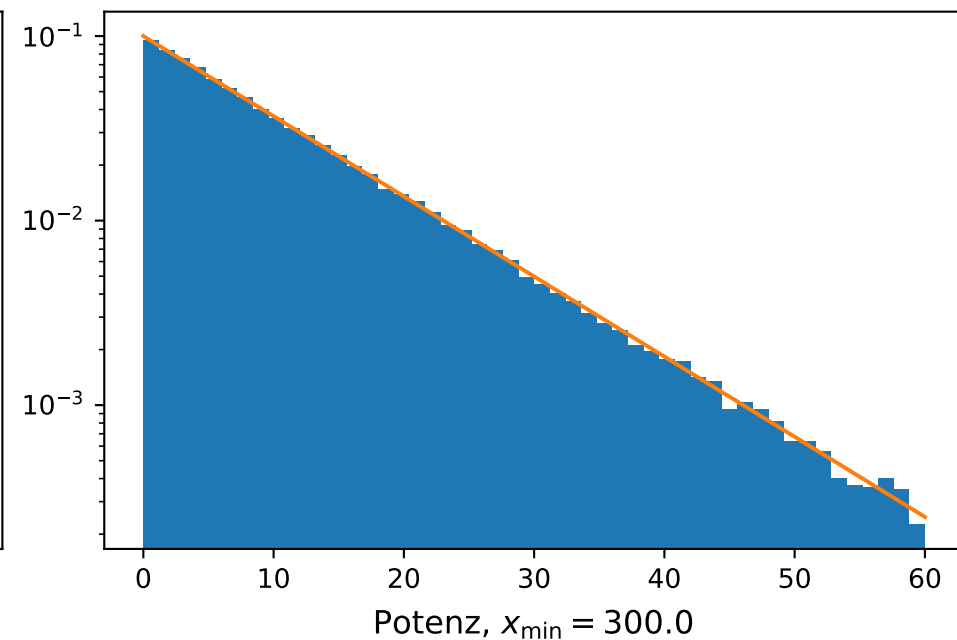
Tests:

m=16: True

std. uniform: True

uniform [-1, 2]: True



Exp.-Verteilung, $\tau = 5$ Exp.-Verteilung, $\tau = 10$ 

Information:

Exercise: Verteilungen

Group name: project_c3

Runtime:

exp: 2.5 ms (ref: 3.2 ms)

power: 4.0 ms (ref: 2.6 ms)

cauchy: 3.3 ms (ref: 3.1 ms)

Tests:

exp mean correct: True

power n correct: True

