

# COMPUTATIONAL PHYSICS

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## Problem set 9

### Problem 1

A linear congruential generator (lcg) for pseudo-random numbers is implemented in source code 1. By using the values  $a = 106$ ,  $c = 1283$  and  $m = 6075$  with the initial states  $I_0 = 110$ ,  $J_0 = 883$  one can check the behavior of this lcg 'by eye'. The idea is to generate (in this case 100)  $x$ ,  $y$  value pairs between 0 and 1, so that the result can be plotted as a distribution of points. This is done in source code 2 with the lcg defined in source code 1 and with the python method `random`, defined in the module `random`, which generates a pseudo-random real number between 0 and 1. The results for both methods are given in figure 1.

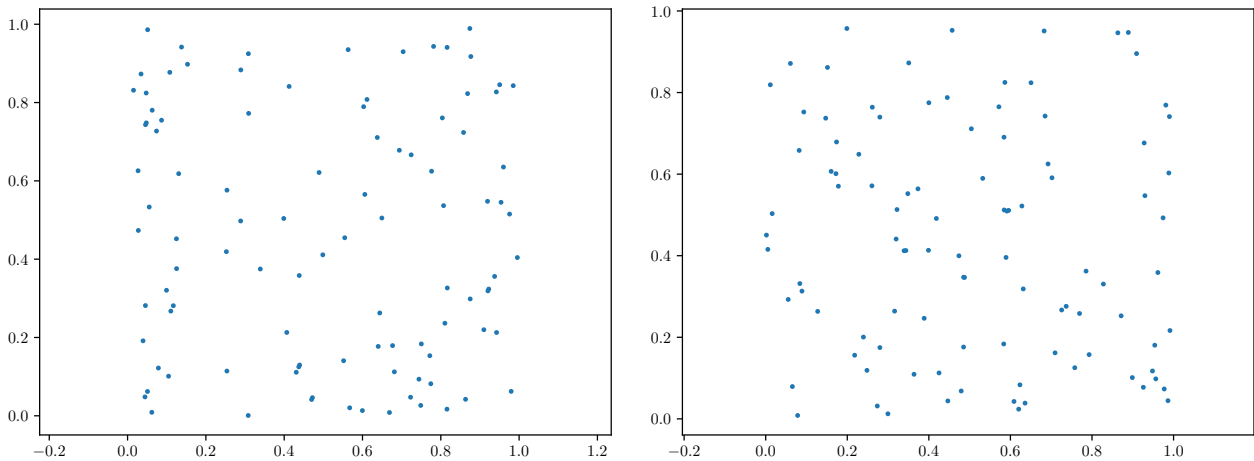


Figure 1: Distribution of 100 points, which  $x$ ,  $y$  values (between 0 and 1) are generated by the lcg generator defined in source code 1 (left) and by the python method `random.random` (right)

Furthermore with an initial state of  $I_0 = 110$ , the  $I_{j+1}, I_j$ -dependence of both random number generators is plotted by source code 3 and shown in figure 2

The expected deterministic character of this plot is not visible.

Lastly the lcg is used to simulate a dice roll. In source code 4 random numbers between 1 and 6 (or rather 0 and 5) are generated 10 times. Then, for 10,000 experiments, the sum of each random numbers is computed and plotted in a histogram. The result is given by figure 3

As one can see, the resulting distribution is approximately given by a gaussian distribution. At this point, it is somewhat odd, that there nearly half of the results are systematically missing. This also happens, by using the `random.random` method and i did not found an explanation for that, just yet.

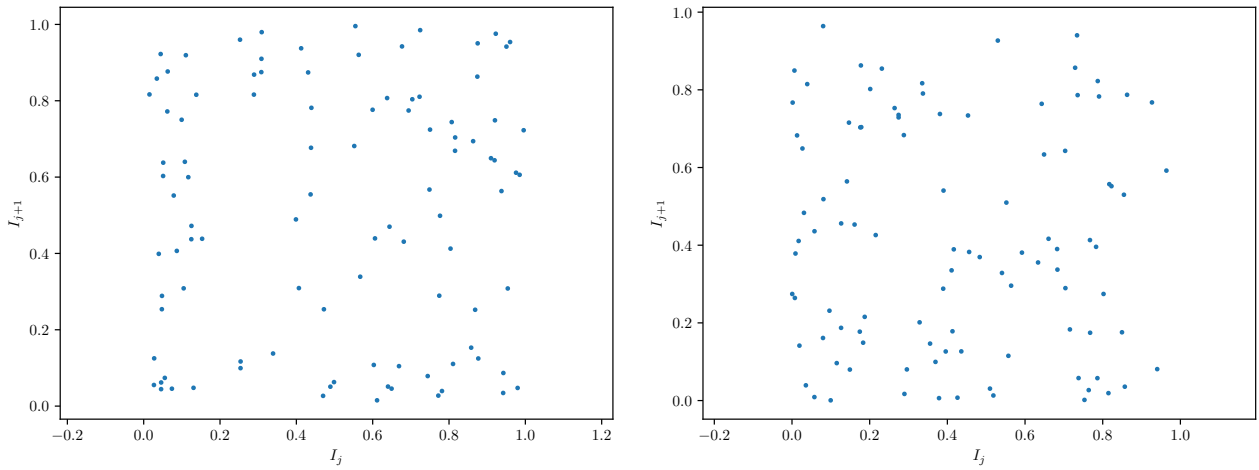


Figure 2:  $I_{j+1}$  in dependence of  $I_j$

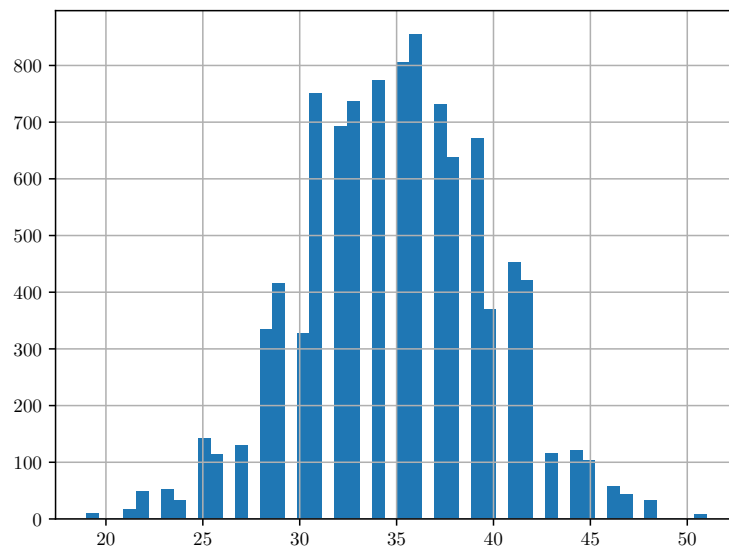


Figure 3: Sum of the results of 10 simulated dice rolls, for 10,000 total experiments. Plotted are the frequencies of each possible result.

```
class rng:
    def __init__(self, seed, a, c, m):
        self.__I = seed
        self.__a = a
        self.__c = c
        self.__m = m

    def next_int(self):
        self.__I = (self.__a * self.__I + self.__c) % self.__m
        return self.__I
```

Source code 1: rng.py

```
import matplotlib.pyplot as plt
import numpy as np
import os
```

```

import random

from rng import rng

figFolderPath = os.path.join(
    os.path.dirname(os.path.realpath(__file__)), 'figures'
)
if not os.path.exists(figFolderPath):
    os.makedirs(figFolderPath)

a, c, m = 106, 1283, 6075
I0, J0 = 110, 883

rng1 = rng(I0, a, c, m)
rng2 = rng(J0, a, c, m)

N = 100
x = np.array([rng1.next_int() / (m - 1) for i in range(N)])
y = np.array([rng2.next_int() / (m - 1) for i in range(N)])

x_ = np.array([random.random() for i in range(N)])
y_ = np.array([random.random() for i in range(N)])

plt.subplots()
plt.grid(False)
plt.axis('equal')
plt.scatter(x, y)

plt.savefig(os.path.join(figFolderPath, 'fig11.pdf'),
            bbox_inches='tight', pad_inches=0.6)
plt.savefig(os.path.join(figFolderPath, 'fig11.pgf'),
            bbox_inches='tight')

plt.subplots()
plt.grid(False)
plt.axis('equal')
plt.scatter(x_, y_)

plt.savefig(os.path.join(figFolderPath, 'fig12.pdf'),
            bbox_inches='tight', pad_inches=0.6)
plt.savefig(os.path.join(figFolderPath, 'fig12.pgf'),
            bbox_inches='tight')

```

Source code 2: problem1\_1.py

```

import matplotlib.pyplot as plt
import numpy as np
import os
import random

```

```

from rng import rng

figFolderPath = os.path.join(
    os.path.dirname(os.path.realpath(__file__)), 'figures'
)
if not os.path.exists(figFolderPath):
    os.makedirs(figFolderPath)

a, c, m = 106, 1283, 6075
I0 = 110

r = rng(I0, a, c, m)

N = 100
I = np.array([r.next_int() / (m - 1) for i in range(N)])
I_ = np.array([random.random() for i in range(N)])

plt.subplots()
plt.grid(False)
plt.axis('equal')
plt.xlabel(r'$I_j$')
plt.ylabel(r'$I_{j+1}$')
plt.scatter(I[:-1], I[1:])

plt.savefig(os.path.join(figFolderPath, 'fig13.pdf'),
            bbox_inches='tight', pad_inches=0.6)
plt.savefig(os.path.join(figFolderPath, 'fig13.pgf'),
            bbox_inches='tight')

plt.subplots()
plt.grid(False)
plt.axis('equal')
plt.xlabel(r'$I_j$')
plt.ylabel(r'$I_{j+1}$')
plt.scatter(I_[:-1], I_[1:])

plt.savefig(os.path.join(figFolderPath, 'fig14.pdf'),
            bbox_inches='tight', pad_inches=0.6)
plt.savefig(os.path.join(figFolderPath, 'fig14.pgf'),
            bbox_inches='tight')

```

Source code 3: problem1\_2.py

```

import matplotlib.pyplot as plt
import numpy as np
import os
import random

from rng import rng

```

```

figFolderPath = os.path.join(
    os.path.dirname(os.path.realpath(__file__)), 'figures'
)
if not os.path.exists(figFolderPath):
    os.makedirs(figFolderPath)

a, c, m = 106, 1283, 6075
I0 = 329

r = rng(I0, a, c, m)

N = 10000
n = np.zeros(N)
for i in range(N):
    n[i] = np.sum([(r.next_int() % 6) + 1 for i in range(10)])

plt.hist(n, bins=50)

plt.savefig(os.path.join(figFolderPath, 'fig15.pdf'),
    bbox_inches='tight', pad_inches=0.6)
plt.savefig(os.path.join(figFolderPath, 'fig15.pgf'),
    bbox_inches='tight')

```

Source code 4: problem1\_3.py

## Problem 2

The Area  $A$  under  $p(x)$  is given by

$$A = \int_0^a p(x) dx = \frac{a^2 b}{2} \quad (1)$$

Under the condition  $A = 1$ , one obtains  $b = \frac{2}{a^2}$ . The maximum of this distribution is  $p_{\max} = \frac{2}{a}$ . With this in mind, one can compute pseudo-random numbers which are distributed accordingly to  $p(x)$ , using the rejection method. This is done in source code 5. The uniform distributed random numbers needed for this are generated by the python method `random.random`. The result is given by the histogram in figure 4. One gets the first fit, which is pretty reasonable with a total number of generated samples of 1,000,000 as used below.

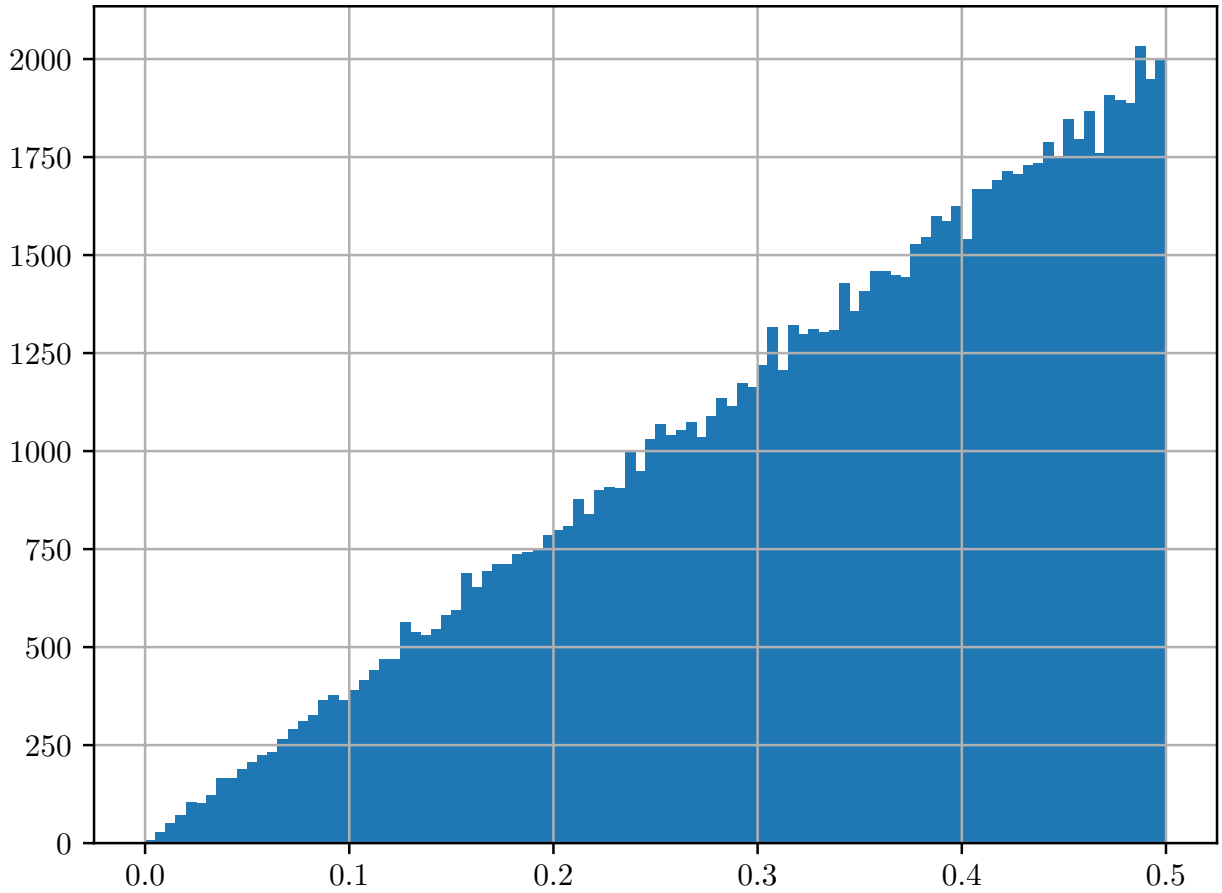


Figure 4: Histogram of 1,000,000 random numbers distributed accordingly to  $p(x)$ , computed using the rejection method.

```

import matplotlib.pyplot as plt
import numpy as np
import os
import random

figFolderPath = os.path.join(
    os.path.dirname(os.path.realpath(__file__)), 'figures'
)
if not os.path.exists(figFolderPath):
    os.makedirs(figFolderPath)

x_max = 0.5
y_max = 2.0 / x_max
b = 2 / x_max**2
def p(x):
    return b * x

N = 1000000
x_samples = np.zeros(N)
for i in range(N):
    while True:
        x = x_max * random.random()
        y = y_max * random.random()
        if y <= p(x):
            break
    x_samples[i] = x

plt.hist(x_samples, 100, range=(0, x_max))

plt.savefig(os.path.join(figFolderPath, 'fig2.pdf'),
            bbox_inches='tight', pad_inches=0.6)
plt.savefig(os.path.join(figFolderPath, 'fig2.pgf'),
            bbox_inches='tight')

```

Source code 5: problem2.py

### Problem 3

The estimation of  $\pi$  is done in source code 6. The idea is to generate points within the square  $[0, 1] \times [0, 1]$  randomly and count the number of accepted samples  $N_{\text{acc}}$ , which are under the function  $f(x) = \sqrt{1 - x^2}$ . Then the ratio  $N_{\text{acc}}/N$ , with the total number  $N$  of the samples, is given by the ratio of the circle quadrant area  $A_{\text{circ}} = \pi/4$  divided by the total area of the square  $A = 1$ . So

$$\pi = 4 \frac{N_{\text{acc}}}{N} \quad (2)$$

The result of this estimation, which is the deviation of the estimations from  $\pi$ , is given in figure 5.

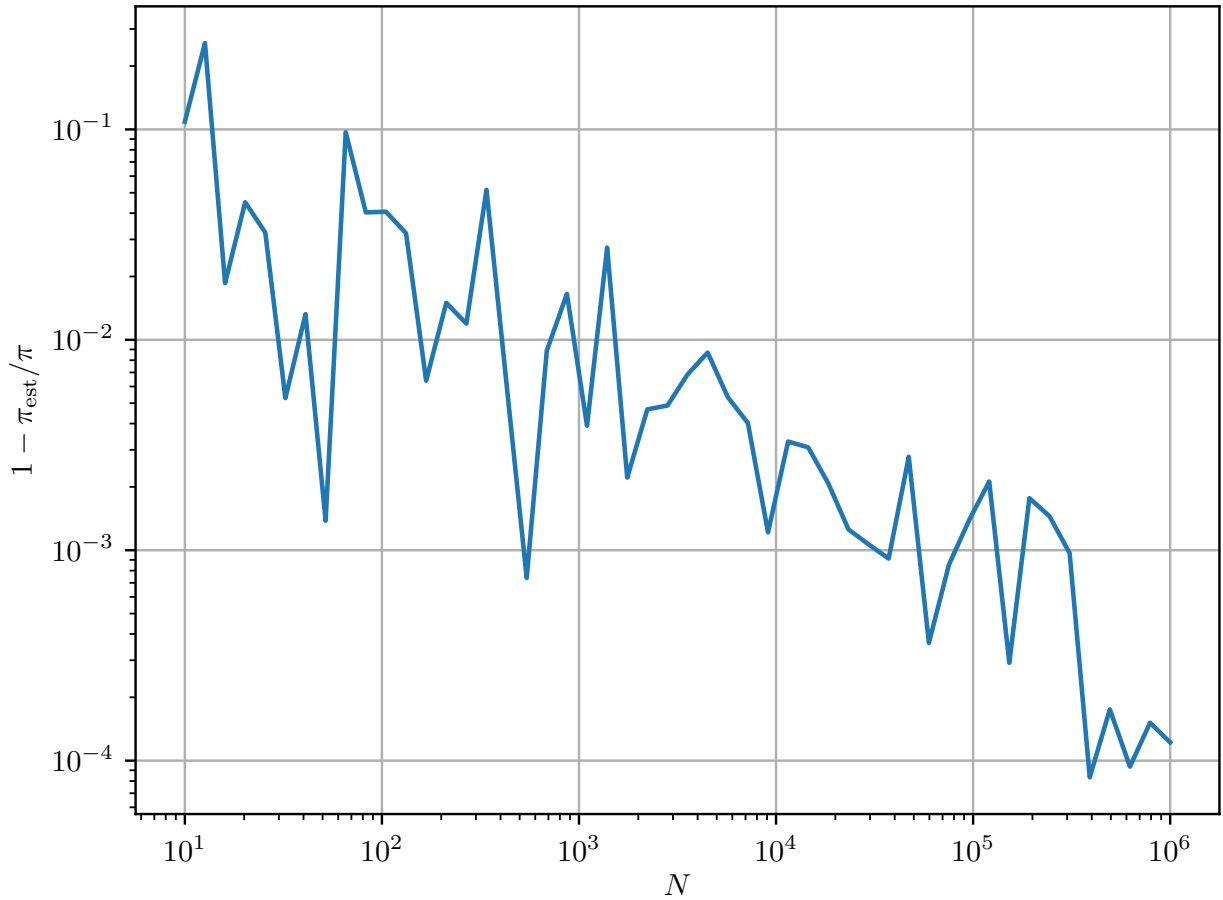


Figure 5: Deviation of the estimation  $\pi_{\text{est}}$  from  $\pi$ , in dependence of the number of samples  $N$ .



```

import matplotlib.pyplot as plt
import numpy as np
from numpy import pi, sqrt
import os
import random

figFolderPath = os.path.join(
    os.path.dirname(os.path.realpath(__file__)), 'figures'
)
if not os.path.exists(figFolderPath):
    os.makedirs(figFolderPath)

def f(x):
    return sqrt(1 - x**2)

def pi_est(N):
    n_acc = 0
    for _ in range(N):
        x = random.random()
        y = random.random()
        if y <= f(x):
            n_acc += 1

    pi_est = 4 * n_acc / N
    return pi_est

N = np.logspace(1, 6, 50)
pi_estimates = np.array([pi_est(int(N_)) for N_ in N])
err = np.array([np.abs(1 - pi_est_ / pi) for pi_est_ in pi_estimates])

for pi_est_ in pi_estimates:
    print(pi_est_)
print()
print(pi)

plt.xlabel(r'$N$')
plt.ylabel(r'$1 - \pi_{\mathrm{est}} / \pi$')
plt.loglog(N, err)

plt.savefig(os.path.join(figFolderPath, 'fig3.pdf'),
            bbox_inches='tight', pad_inches=0.6)
plt.savefig(os.path.join(figFolderPath, 'fig3.pgf'),
            bbox_inches='tight')

```

Source code 6: problem3.py