

# Tausworthe PRN report

---

**Student:** Andrei Goncharov (agoncharov6@gatech.edu)

## Problem statement

Implement the Tausworthe pseudo-random number generator. Perform a decent number of statistical tests on the generator to see that it gives PRN's that are approximately i.i.d. Uniform(0,1). Plot adjacent PRN's ( $U_i, U_{i+1}$ ),  $i = 1, 2, \dots$ , on the unit square to see if there are any patterns. Generate a few Nor(0,1) deviates (any way you want) using Unif(0,1)'s from the Tausworthe generator.

## Approach

1. Implement a base class for the Tausworthe generator.
2. Implement a simple CLI to interact with the class
3. Add commands to the CLI to:
  1. Generate a sequence of PRNs (pseudo-random numbers) and print them.
  2. Generate PRNs and test them. Print the graphs. Generate Nor(0, 1) deviates.

## CLI

Using [python-nubia](#) create a simple CLI. Read more about it in the README.

CLI has limitations! It currently does not do any input validation. If you enter bad data, you are on your own!

## Tausworthe generator class

Accepts initial `seed`, `q`, `r`, `l`. Exposes a single method `next` that returns a new `U(i)`. Internally, keeps the current number of `q` bits as `seq`. To generate a new PRN it appends `l` new bits to `seq` and pop `l` first bits. `seq` is initialized to `seed`.

Every new bit is calculated by the algorithm:  $B(i) = B(i - q) \text{ xor } B(i - r)$ .

Every new PRN requires `l` new bits and is calculated as `l` converted from binary to decimal divided by  $2^l$ .

## Statistical tests

We will perform a goodness-of-fit test (with  $\alpha = 0.05$  and 101 bins) and an independence test (runs "Up and Down") (with  $\alpha = 0.05$ ).

We use 101 bins so our chi squared distribution has 100 degrees of freedom.

## Choosing constants

Tausworthe algorithm uses several constants:

- `q` - big number of offset bits
- `r` - little number of offset bits

- $l$  - little number of offset bits

$q$  must be as big as possible as we generally want as big as possible cycles for our PRN generators. So we set  $q$  to 31 bits by default. It gives us a cycle of  $2^{31}-1$  which is the biggest any 32-bit machine can handle.

$r$  and  $l$  can vary. Due to the limited scope of the project, we set  $l$  to 20. It makes the generator produce  $2^{20}$  different PRNs which sounds like a reasonable number.

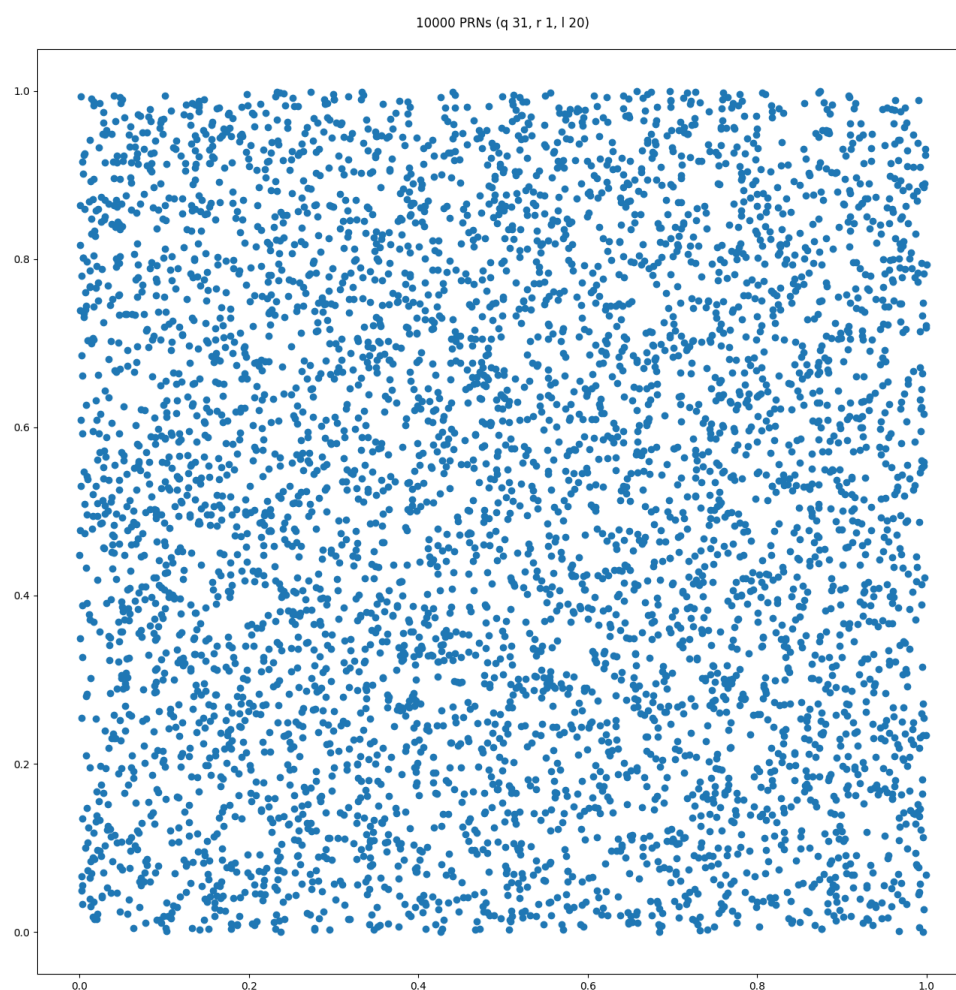
For  $r$  we will perform a few tests to see if value of  $r$  significantly affects the result.

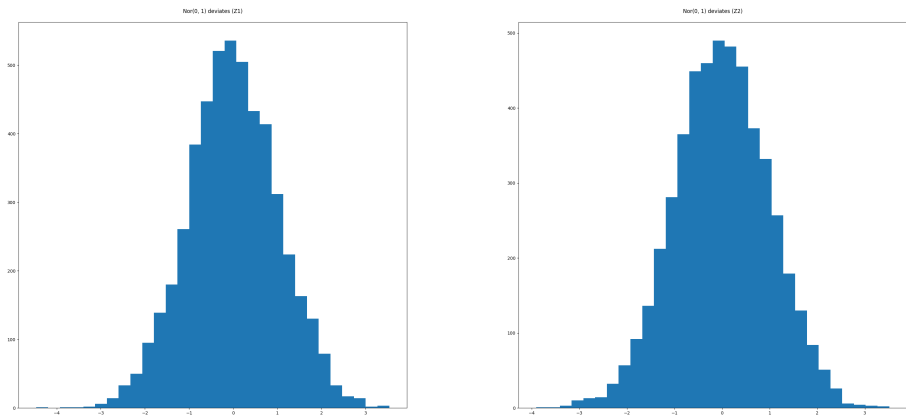
## Testing $r$

Using 42424242424242 as our seed. We will generate 10000 numbers. Based on the constants listed above, our reference (threshold) Z-score is 1.96. Reference chi squared value is 124.342.

$r = 1$

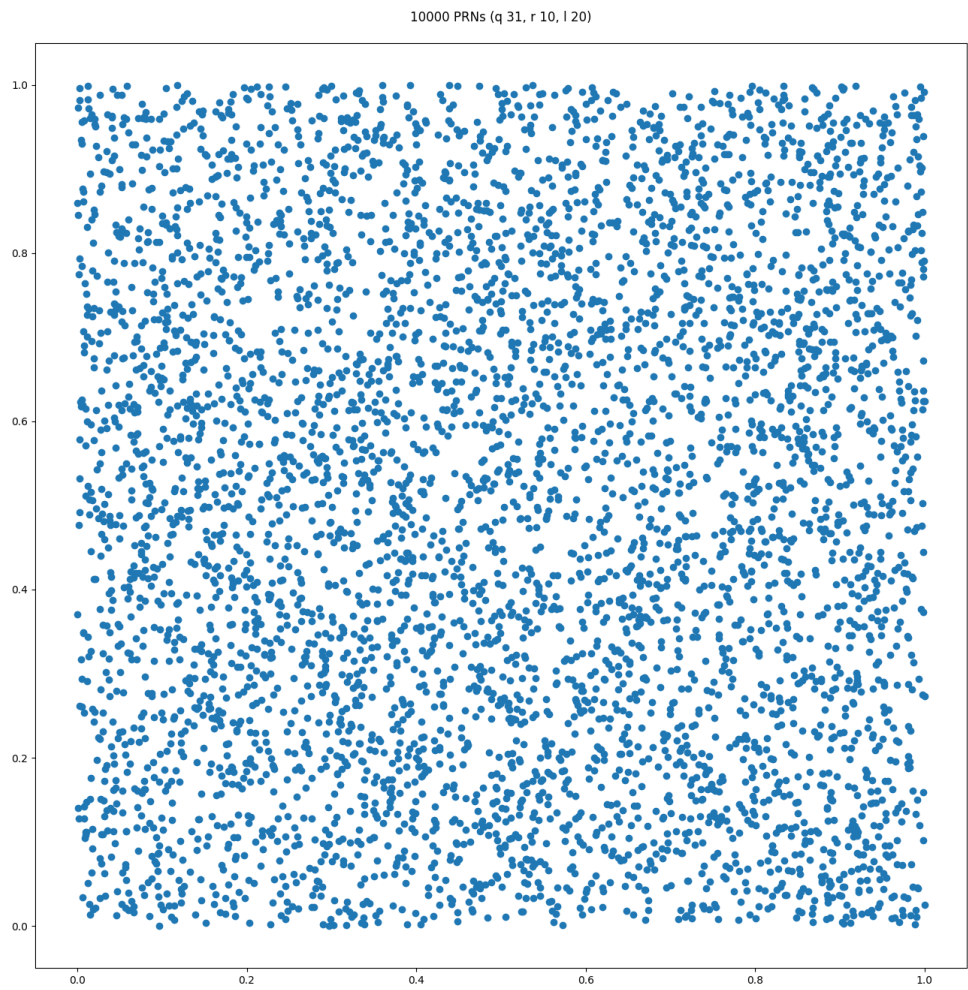
Chi<sup>2</sup> = 97.7578. Number of runs = 6708.

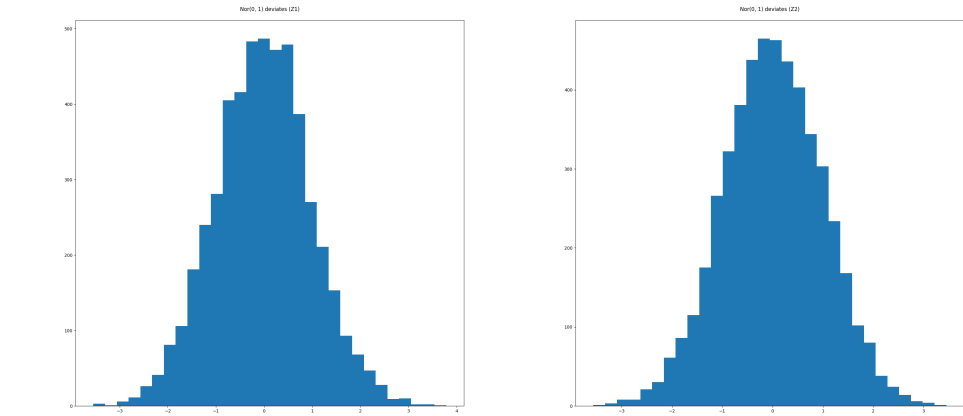




$r = 10$

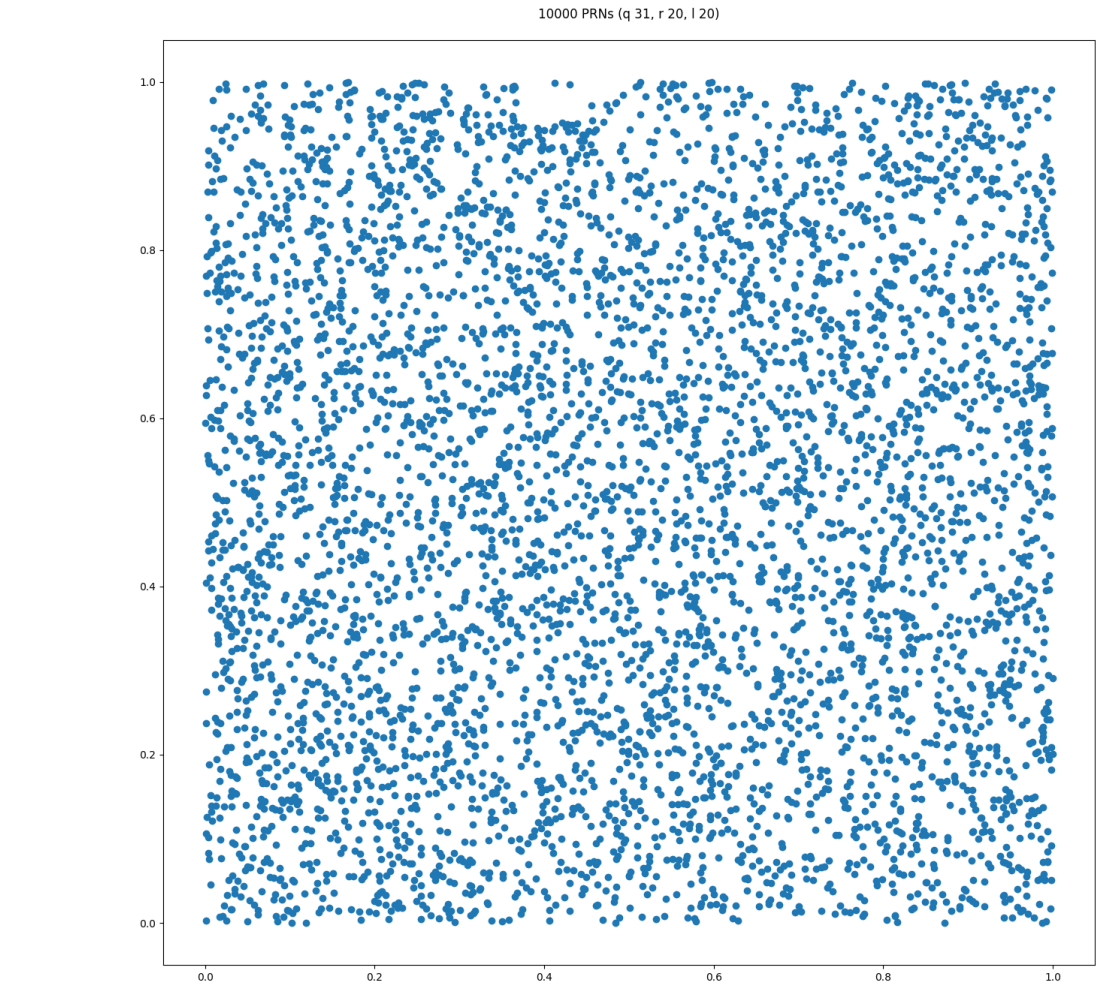
$\chi^2 = 77.75$ . Number of runs = 6659.

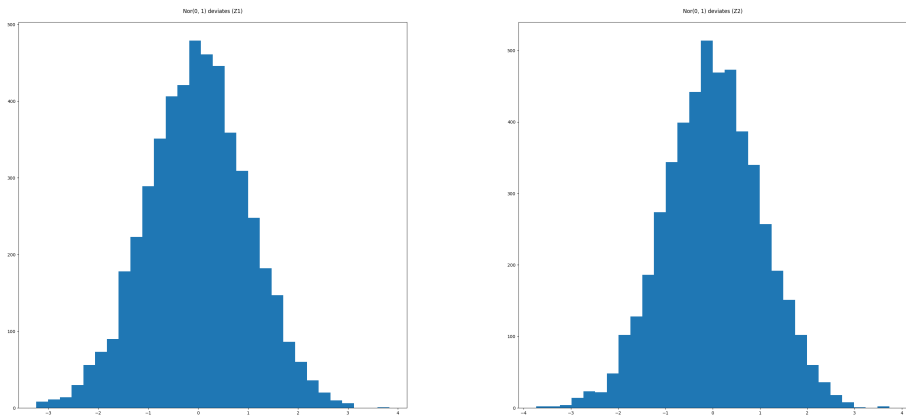




r = 20

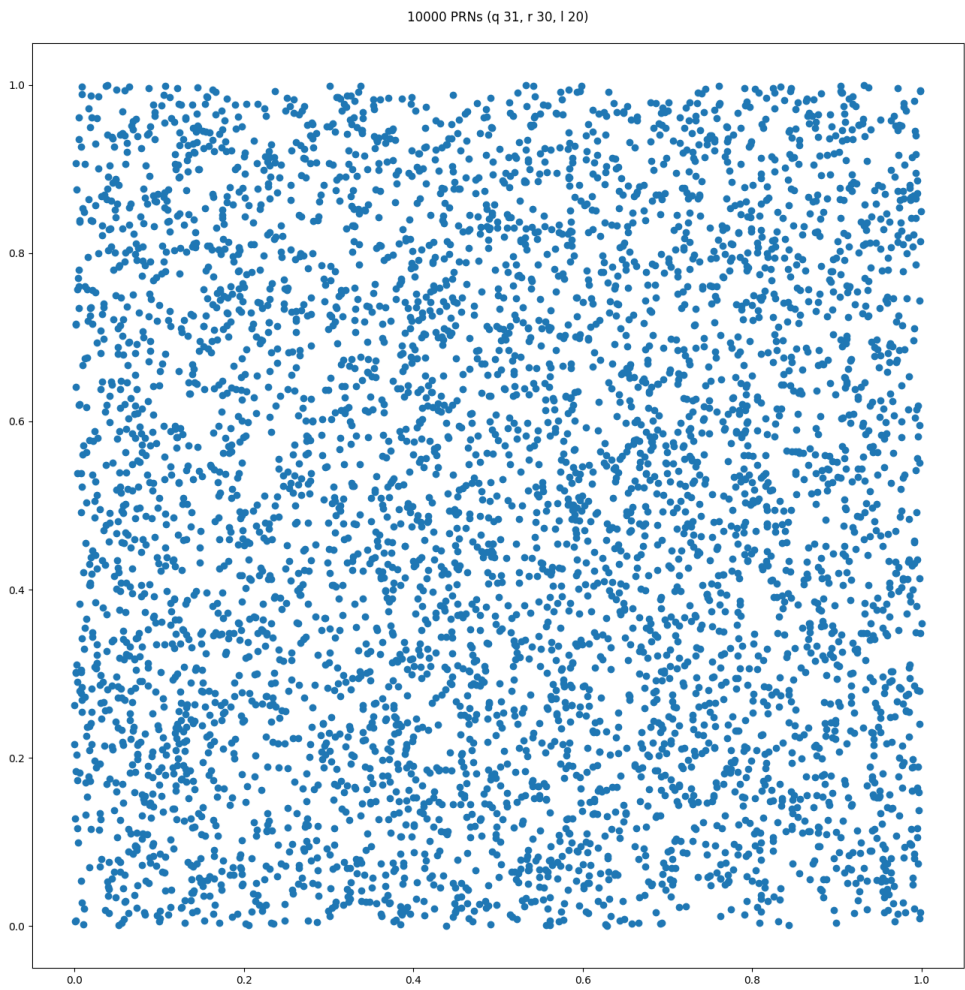
Chi^2 = 100.6868. Number of runs = 6612.

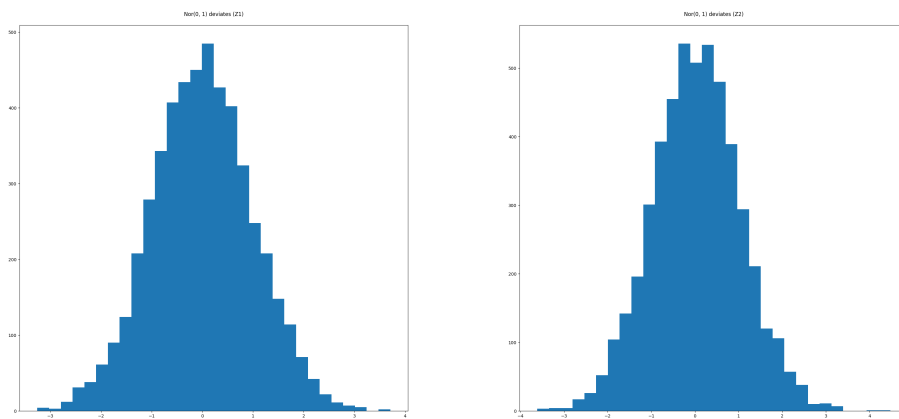




**r = 30**

Chi^2 = **104.89**. Number of runs = **6643**.





## Summary

As we can see, our Tausworthe generator produces PRNs that pass goodness-of-fit and independence tests, thus we conclude that the implementation is correct. We can also notice that different values of  $r$  did not affect the out come significantly, thus we conclude that  $r$  does not affect generator properties and can be anything between 1 and  $q - 1$ . Finally, we can clearly see from the graphs that we can generate Nor(0, 1) deviates using Box-Muller method from the PRNs produces by our generator.

## References

1. [https://studio.edx.org/assets/courseware/v1/f33427e02d57018e66a65baa3c8305c5/asset-v1:GTx+ISYE6644x+2T2019+type@asset+block/Module06-RandomNumberGenerationSlides\\_180526.pdf](https://studio.edx.org/assets/courseware/v1/f33427e02d57018e66a65baa3c8305c5/asset-v1:GTx+ISYE6644x+2T2019+type@asset+block/Module06-RandomNumberGenerationSlides_180526.pdf)