

- Applications of Monte Carlo simulations.
- Random number generators.

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- Coin toss: the only prediction about the outcome: 50% of the events will end up as tail being up.
- Dice: In a large number of throws, the probability of getting a given face is  $\frac{1}{6}$ .

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- Often need a program that generates a random variable with a given probability distribution.

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- It implies that no matter which pseudo random number generator you use – it will always repeat itself (period of the generator).

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- Feedback shift register methods

## Linear congruent method for RNG

Generates a pseudo random sequence of numbers  $\{x_1, x_2, \dots, x_k\}$  of length  $M$  over the interval  $[0, M - 1]$ :

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Note that

$$\text{mod } (b, M) = b - \text{int}(b/M) * M$$

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$$x_2 = 4$$

$$x_3 = 8$$

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interval:  $0 - 8$  i.e.  $[0, M - 1]$

Period:  $9$  i.e.  $M$  numbers (then repeat).

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- For  $c = 0$  called "Multiplicative congruential generator".

## Random numbers on interval $[A, B]$

- Scale results from  $x_i$  on  $[0, M - 1]$  to  $y_i$  on  $[0, 1]$ .

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- Scale results from  $y_i$  on  $[0, 1]$  to  $z_i$  on  $[A, B]$

$$z_i = A + (B - A) * y_i$$

## Feedback shift register generator

Simple shift register where the vacated bit is filled with the exclusive-or of two other bits in the shift register.

## Feedback shift register generator

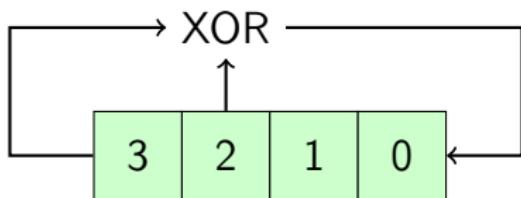
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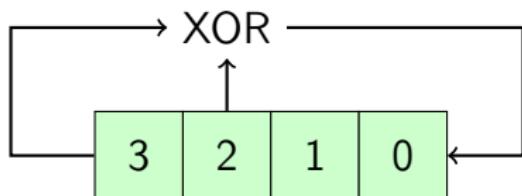
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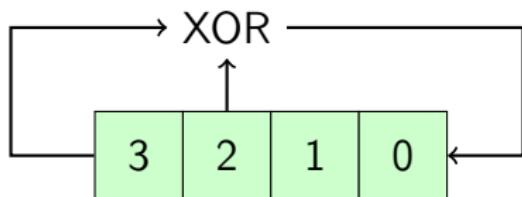


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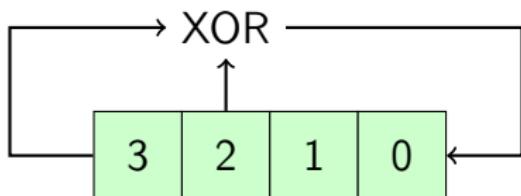


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4 bit shift-register pseudorandom number generator:



- Bits 3 and 2 are combined by exclusive-or.
- The register is shifted 1 step to the left.
- The result of the exclusive-or is entered into bit 0.

# 4bit shift register PRNG

Here is the pattern of bits, starting with 0001:

0001

0010

0100

1001

0011

0110

1101

1010

0101

1011

0111

1111

1110

1100

1000

0001

- Most commonly used is Mersenne Twister (which is a generalized feedback shift register method).

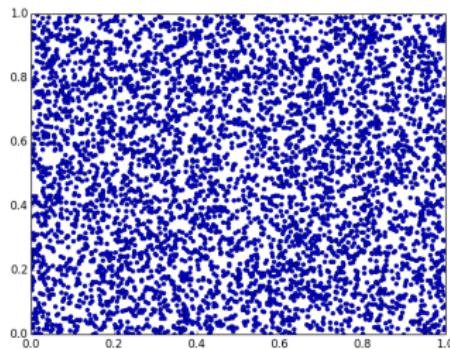
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- Implemented in numpy.

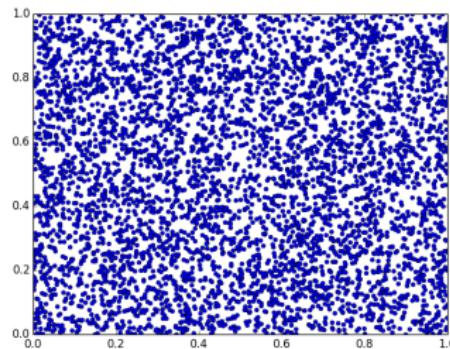
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- 2D plot,  $x_i$  and  $y_i$  from two random sequences (parking lot test).



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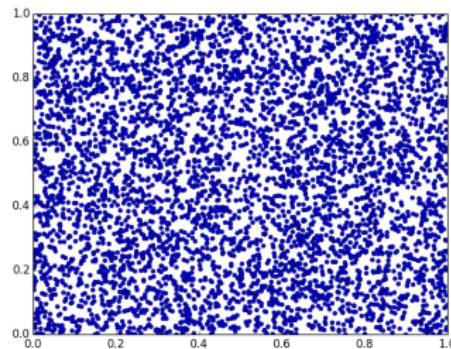
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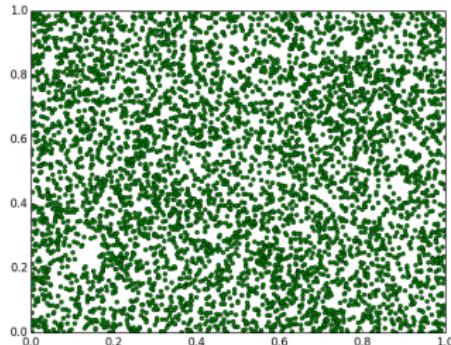
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- Plot 3D figure  $(x_i, y_i, z_i)$
- Plot correlation  $(x_i, x_{i+k})$



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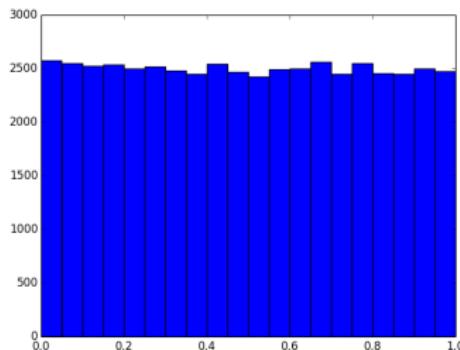
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- Near neighbour correlation:

$$\frac{1}{N} \sum_{i=1}^N x_i x_{i+k} \approx \frac{1}{4}$$

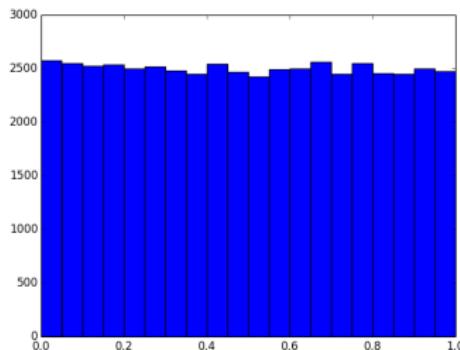
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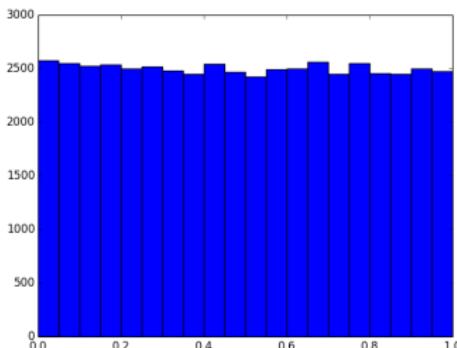
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- near neighbor correlation: (50000 random numbers)  
= 0.2478

Good test suites exist – TestU01 – which can be used to uncover problems in random number generators.

Dont try to invent your own random number generator – unless you know what you are doing. This is very tricky business!!!