Lecture 7: Basic Sampling

Professor Ilias Bilionis

Pseudo-random number generators



Pseudo-random number generators

- Computers are deterministic machines and therefore they cannot generate completely random numbers?
- Idea: Are there deterministic sequences of numbers that look random?
- Pseudo-random number generators do exactly that.
- We use statistical tests to see how good they are.

Pseudo-random number generators

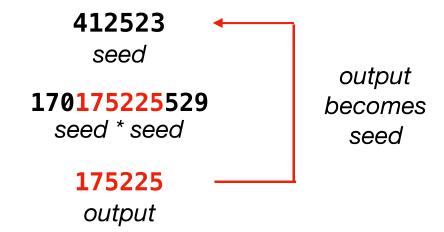
How do you generate a uniform random number?



John von Neumann. (Los Alamos)

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The middle-square method



The first, but it doesn't pass all statistical tests.

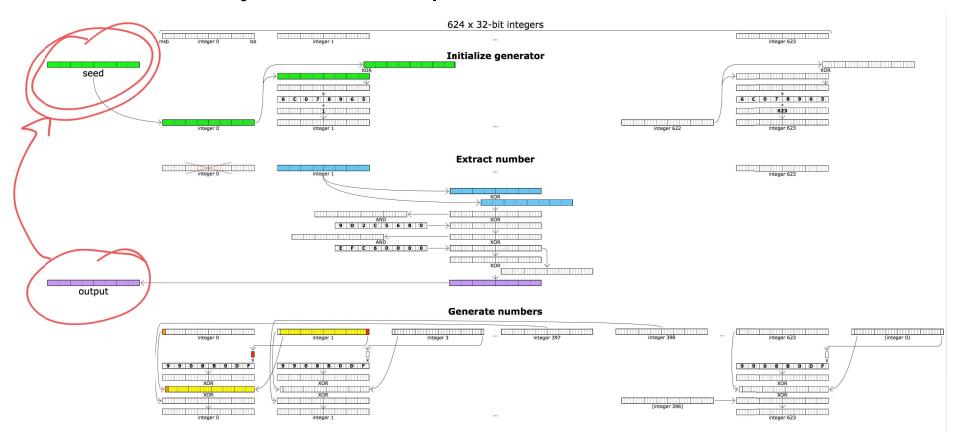
Linear congruential generators

Seed x_0

$$x_{i+1} = (0x_i + 0) \mod m$$

Mersenne Twister PRNG

- This is what is inside numpy.random.
- Details beyond the scope of this class.



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Sampling the uniform

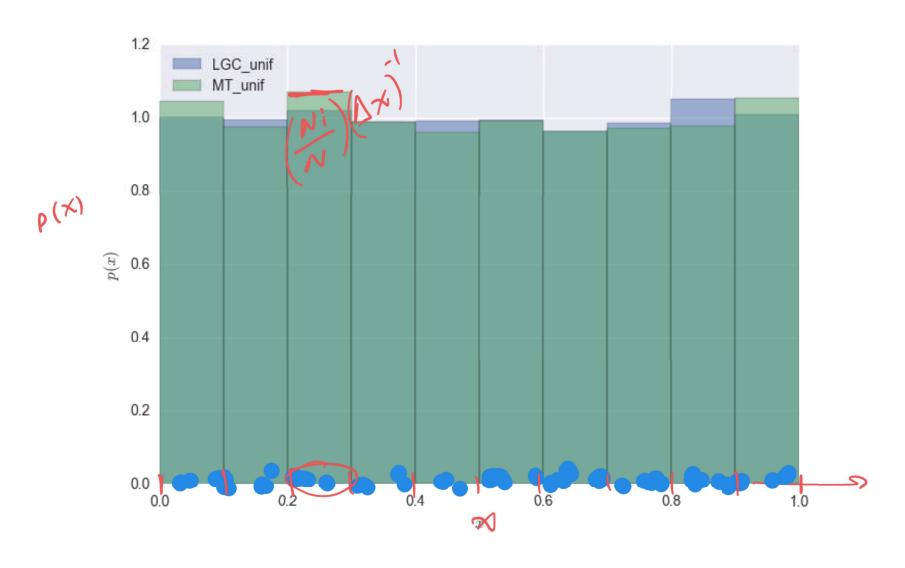


PRNG to uniform

- PRNG's generate random integers from 0 to m.
- How can we get samples from the uniform?
- Step 1: Sample a random integer d.
- Step 2: Set:

$$x = \frac{d}{m}$$

PRNG to Uniform



How do we know that the samples are indeed uniform? *\frac{\sqrt{\lambda} \cdot \cdot \lambda \lambda

$$F(x) = P[X \in x] = x$$

We can compare the empirical CDF with the ideal CDF.

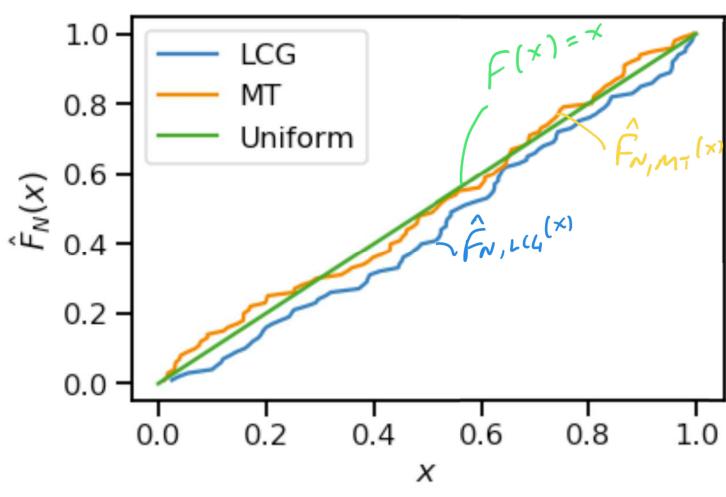
But what is the empirical CDF of a bunch of samples $x_{1\cdot N}$?

It is defined as follows:

$$\hat{F}_N(x) = \frac{\text{number of elements in sample } \leq \underline{x}}{\underline{N}}$$



How do we know that the samples are indeed uniform?





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Sampling the categorical



Example: Sampling from the Bernoulli distribution

$$X \sim \text{Bernoulli}(\theta); X = \begin{cases} 1, & \text{w/pr. } \theta \\ 0, & \text{otherwise} \end{cases}$$

To sample from it, we do the following steps:



Sampling discrete distributions

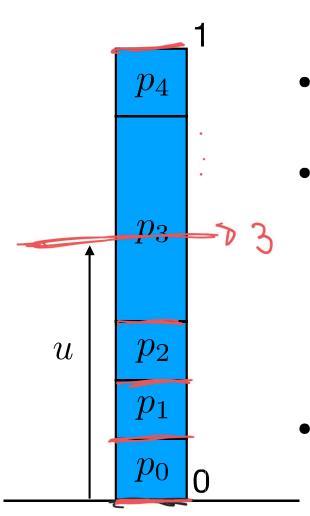
 Consider a generic discrete random variable taking different values, with probability:

$$p(X=k)=p_{k}.$$

$$\sum_{k=1,\,ul\,pr.\,P_{k-1}} \sum_{k=1,\,ul\,pr.\,P_{k-1}} \sum_{k=1,\,ul\,p$$



Sampling Discrete Distributions



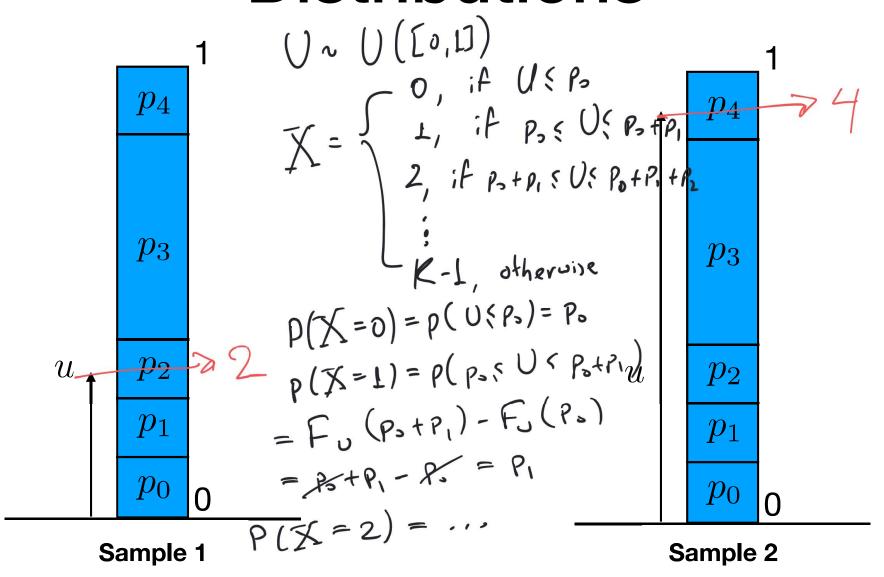
• Draw a uniform number $u \sim (0,1)$

• Find *j* such that:

$$\sum_{k=0}^{j-1} p_k \le u < \sum_{k=0}^{j} p_k$$

• *j* is your sample

Sampling Discrete Distributions



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Inverse sampling



Inverse Sampling

- Consider an arbitrary univariate continuous random variable X with CDF F(x), How do you sample from it?
- Draw a uniform number u. ~ U(L), I)
- Set:

$$x = F_1^{-1}(u)$$
inverse of the COF

and you get your sample!

Why does inverse sampling work?

- Let $U \sim U([0,1])$ be a uniform random variable.
- For any CDF F(x) define the random variable:

$$X = F^{-1}(U)$$

The CDF of X is:

$$p(X \le x) = p \left(F'(U) \le x \right) = p \left(F(F'(U)) \le F(x) \right)$$

$$= p \left(U \le F(x) \right) = F_U \left(F(x) \right) = F(x)$$



Example: The exponential distribution

• Take an exponential random variable as an example:

$$X \sim \text{Exp}(r)$$

The CDF is:

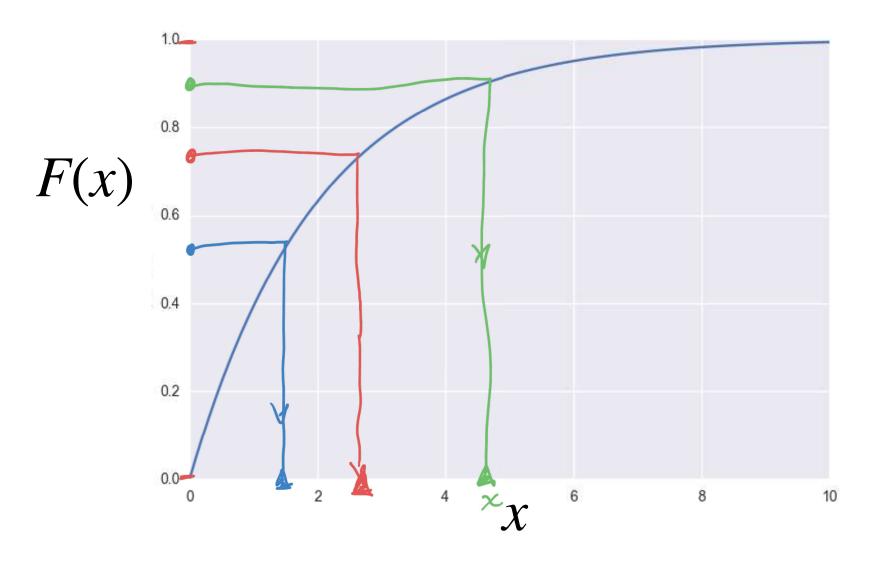
$$F(x) = 1 - e^{-rx}$$

The inverse of the CDF is:

$$F^{-1}(u) = -\frac{\ln(1-u)}{r}$$



The Exponential Distribution



Inverse Sampling for Exponential

