

# ECE 302: Lecture 4.9 Generating Random Numbers

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# Outline

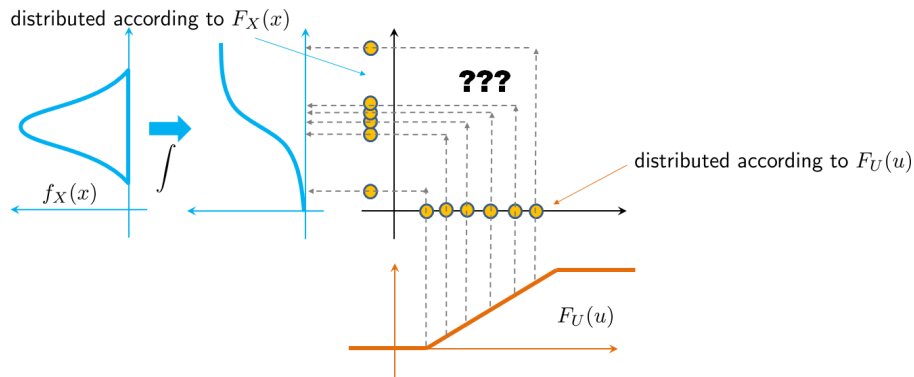
## Overall schedule:

- Continuous random variables, PDF
- CDF
- Expectation
- Mean, mode, median
- Common random variables
  - Uniform
  - Exponential
  - Gaussian
- Transformation of random variables
- **How to generate random numbers**

## Today's lecture:

- General principle
- Examples

# Principle



**Figure:** Generating random number according to a known CDF. The idea is to first generate a uniform(0,1) random variable, and then do an inverse mapping  $F_X^{-1}$ .

# Main theorem

## Theorem

The transformation is

$$g(u) = F_X^{-1}(u). \quad (1)$$

That is, if  $g = F_X^{-1}$ , then  $g(U)$  will be distributed according to  $f_X$  (or  $F_X$ ).

First, if  $U \sim \text{Uniform}(0, 1)$ , then  $f_U(u) = 1$  for  $0 \leq u \leq 1$  and so

$$F_U(u) = \int_{-\infty}^u f_U(u) du = u, \quad 0 \leq u \leq 1.$$

Let  $g = F_X^{-1}$  and define  $Y = g(U)$ . Then the CDF of  $Y$  is

$$\begin{aligned} F_Y(y) &= \mathbb{P}[Y \leq y] \\ &= \mathbb{P}[g(U) \leq y] \\ &= \mathbb{P}[F_X^{-1}(U) \leq y] \\ &= \mathbb{P}[U \leq F_X(y)] = F_X(y). \end{aligned}$$

## Idea

**How to generate random numbers from an arbitrary distribution  $F_X$ ?**

- Step 1: Generate a random number

$$U \sim \text{Uniform}(0, 1).$$

- Step 2: Let

$$Y = F_X^{-1}(U). \quad (2)$$

Then the distribution of  $Y$  is  $F_X$ .

**MATLAB:** Suppose we are told that  $F_X^{-1}(U) = -\frac{1}{\lambda} \log(1 - U)$ .

- Step 1: `U = rand(10000,1)`
- Step 2: `Y = -(1/lambda)*log(1-U)`

(Similar commands can be found in Python.)

## Examples

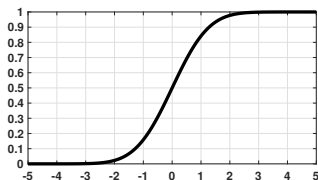
**Example 1.** How to generate, from uniform random numbers to, **Gaussian** random numbers with mean  $\mu$ , and variance  $\sigma^2$ ?

First of all, we generate  $U \sim \text{Uniform}(0, 1)$ . The CDF of the ideal distribution is

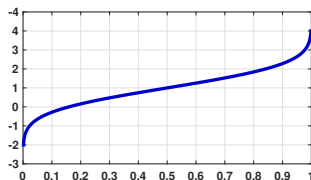
$$F_X(x) = \Phi\left(\frac{x - \mu}{\sigma}\right).$$

Therefore, the transformation  $g$  is

$$g(U) = F_X^{-1}(U) = \sigma\Phi^{-1}(U) + \mu.$$



(a)  $F_X(x)$



(b)  $g(U)$

## Examples

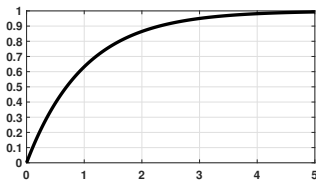
**Example 2.** How to generate, from uniform random numbers to, **exponential** random numbers with parameter  $\lambda$ ?

First of all, we generate  $U \sim \text{Uniform}(0, 1)$ . The CDF of the ideal distribution is

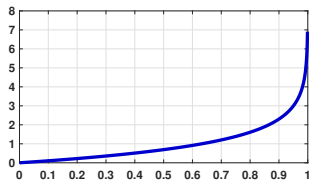
$$F_X(x) = 1 - e^{-\lambda x}.$$

Therefore, the transformation  $g$  is

$$g(U) = F_X^{-1}(U) = -\frac{1}{\lambda} \log(1 - U).$$



(a)  $F_X(x)$



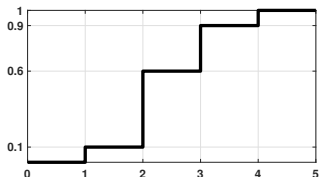
(b)  $g(U)$

## Examples

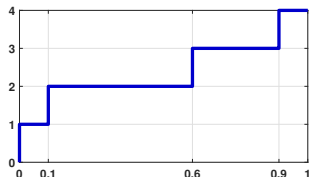
**Example 3.** How to generate, from uniform random numbers to, 4 numbers 1, 2, 3, 4 according to the histogram [0.1 0.5 0.3 0.1].

First of all, we generate  $U \sim \text{Uniform}(0, 1)$ . The CDF of the ideal distribution is

$$F_X(x) = \begin{cases} 0.1, & x = 1, \\ 0.1 + 0.5 = 0.6, & x = 2, \\ 0.1 + 0.5 + 0.3 = 0.9, & x = 3, \\ 0.1 + 0.5 + 0.3 + 0.1 = 1.0, & x = 4. \end{cases}$$



(a)  $F_X(x)$



(b)  $g(U)$



## Examples

This CDF is not invertible. However, we can still define the “inverse” mapping as

$$g(U) = F_X^{-1}(U) \\ = \begin{cases} 1, & 0.0 \leq U \leq 0.1, \\ 2, & 0.1 \leq U \leq 0.6, \\ 3, & 0.6 \leq U \leq 0.9, \\ 4, & 0.9 \leq U \leq 1.0. \end{cases}$$

**How to generate random numbers from an arbitrary distribution  $F_X$ ?**

- Step 1: Generate a random number

$$U \sim \text{Uniform}(0, 1).$$

- Step 2: Let

$$Y = F_X^{-1}(U). \tag{3}$$

Then the distribution of  $Y$  is  $F_X$ .

**Questions?**