

Lecture A5. Simulation 2

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- 1 I. Random uniform number
- 2 II. Inverse transform method
- 3 III. Various random numbers

$U(0,1)$

↓ $\times 2$

$U(0,2)$

↓ -1

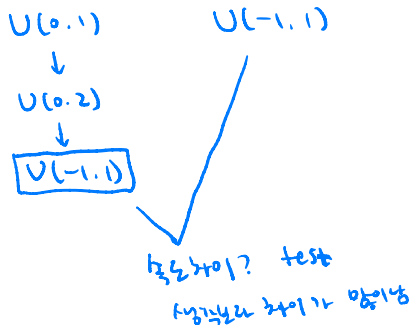
$U(-1,1)$ ■

$U(0,1)$?
[other dist? $\text{Exp}(\lambda)$
R functions for generating
various random numbers.

I. Random uniform number

Recap

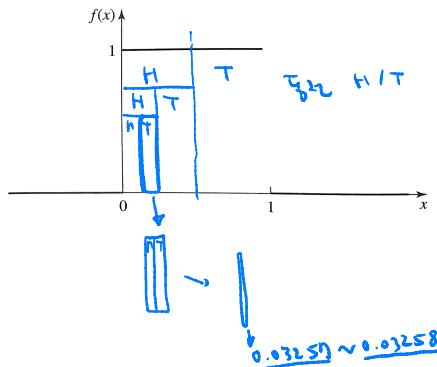
- In the previous simulation lecture, random numbers that follows $U(-1, 1)$ were the initial components of the simulation process for estimating π .
- Since a random variable that follows $U(-1, 1)$ is merely a linear transformation of $U(0, 1)$, we will discuss the generation process for $U(0, 1)$.



$U(0, 1)$

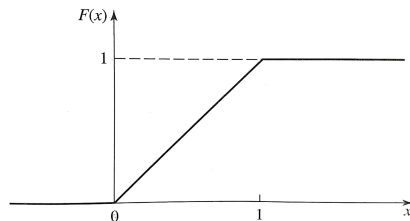
• pdf

$$\text{pdf } f(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



• cdf

$$\text{cdf } F(x) = \begin{cases} 0 & \text{if } x < 0 \\ x & \text{if } 0 \leq x < 1 \\ 1 & \text{if } x \geq 1 \end{cases}$$



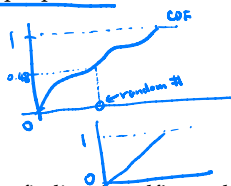
- 175 $4:02:03.356$ $U(0.1)$

II. Inverse transform method

Motivation for random number generation from a general cdf.

- For a continuous random variable X , its cdf has following properties.

- Its lower limit is always 0. ✓
- Its upper limit is always 1. ✓
- The function is always monotonically non-decreasing. ✓



- Discussion

- From the property 3 above, a cdf is one-to-one function.
- Since one-to-one, the cdf has an inverse function. It means that finding the cdf's y -value automatically gives the function's x -value.
- The function's y value is in the bounded interval $[0, 1]$

$$0 \leq F(x) \leq 1. \quad \textcircled{1} \quad u = F(x) \Rightarrow x = F^{-1}(u)$$

- Motivated by the above points of 2 & 3, one can simply 1) find u from $U(0, 1)$, and then 2) take its inverse value with respect to the cdf.

$$u = F(x) \Rightarrow x = F^{-1}(u).$$

then x is rand number whose CDF is $F(\cdot)$

Inverse transform method

★ Theorem 1 (inverse transform method)

If X is a continuous random variable with cdf $F(x)$, then the random variable's CDF $F(X) \sim U(0, 1)$.

Remark 1

The above theorem suggests a way to generate realizations of the random variable X . Namely,

- 1 Pick u from $U(0, 1)$ ✓
- 2 Solve $u = F(x)$ for x , or $x = F^{-1}(u)$. ✓
- 3 Then, x is a random number from the random variable with cdf $F(x)$ ✓

Exponential random numbers

Remark 2

For example, we want to find a x from $X \sim \text{exp}(5)$ and we picked $u = 0.3$ from $U(0, 1)$, then what is the random number x that follows $\text{exp}(5)$?

① $u = 0.3$ ✓ *Fun*

② $u = 1 - e^{-5x} \Rightarrow 1 - u = e^{-5x} \Rightarrow \log(0.7) = -5x \Rightarrow x = \frac{-\log(0.7)}{5}$

③ $x = \frac{-\log(0.7)}{5}$ ✓

Exercise 1

→ function that generates $U(0,1)$

Using runif() function in R, complete the following code block that generates 1,000 random numbers that follow exp(3).

1: $N \leftarrow 1000$ ✓

2: $u \leftarrow \text{runif}(N)$ ✓

3: $x \leftarrow$ (complete here) ✓

4: $\text{head}(x)$

- Uniform random number is indeed the building block for all random numbers!
- What about a random number from a discrete distribution? It's easy.

Random number for discrete distribution

- Suppose a discrete r.v. X has the distribution of the following.

x	1	2	3	4
$\mathbb{P}(X = x)$.1	0	.4	.5
$\mathbb{P}(X \leq x)$.1	.1	.5	1.0

pmf
cdf

- The process is the same. First to pick a u from $U(0, 1)$. Next,
 - if $u \leq .1$, then let $x = 1$.
 - if $.1 < u \leq .5$, then let $x = 3$.
 - if $.5 < u$, then let $x = 4$.
- x is a random number for X .

III. Various random numbers

Using built-in function

- Most programming languages provide built-in random number generator.
- R does so as well with functions whose prefix r- such as runif(), rnorm(), rexp(), rpois(), and so on.
- Code in help(runif) in console opens helper as below.

R: The Uniform Distribution

Uniform (stats) R Documentation

The Uniform Distribution

Description

These functions provide information about the uniform distribution on the interval from `min` to `max`. `dunif` gives the density, `punif` gives the distribution function, `qunif` gives the quantile function and `runif` generates random deviates.

Usage

```
dunif(x, min = 0, max = 1, log = FALSE)
punif(q, min = 0, max = 1, lower.tail = TRUE, log.p = FALSE)
qunif(p, min = 0, max = 1, lower.tail = TRUE, log.p = FALSE)
runif(n, min = 0, max = 1)
```

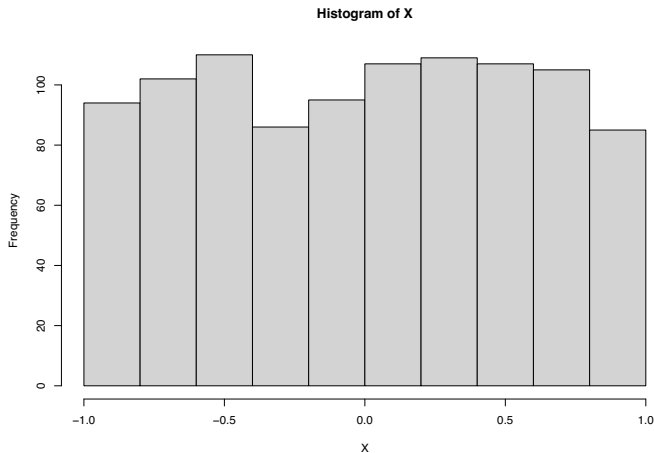
Arguments

<code>x, q</code>	vector of quantiles.
<code>p</code>	vector of probabilities.
<code>n</code>	number of observations. If <code>length(n) > 1</code> , the length is taken to be the number required.
<code>min, max</code>	lower and upper limits of the distribution. Must be finite.
<code>log, log.p</code>	logical; if TRUE, probabilities <code>p</code> are given as <code>log(p)</code> .
<code>lower.tail</code>	logical; if TRUE (default), probabilities are $P(X \leq x)$, otherwise, $P(X > x)$.

Handwritten note: $U(a,b) : \text{runif}(n, a, b)$

Uniform random numbers

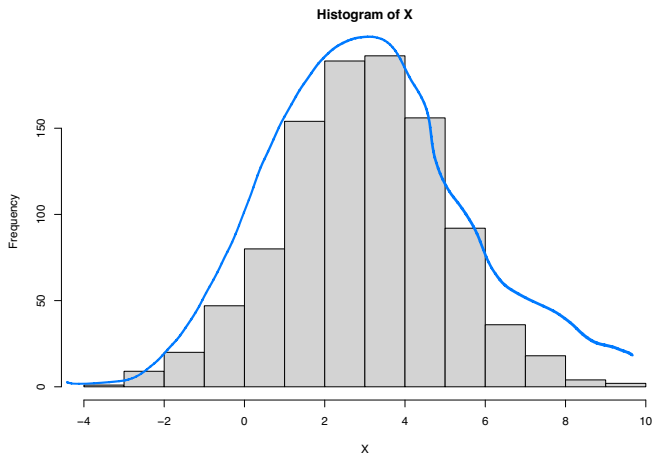
```
X <- runif(n=1000, min=-1, max=1)
hist(X)
```



Normal random numbers

```
X <- rnorm(n=1000, mean=3, sd=2)  
hist(X)
```

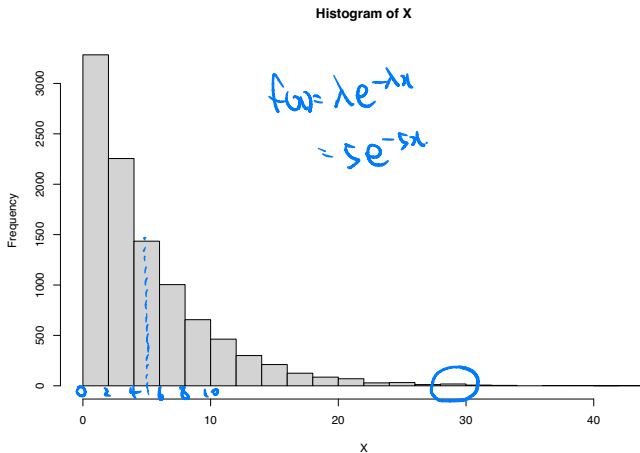
m *6*



Exponential random numbers

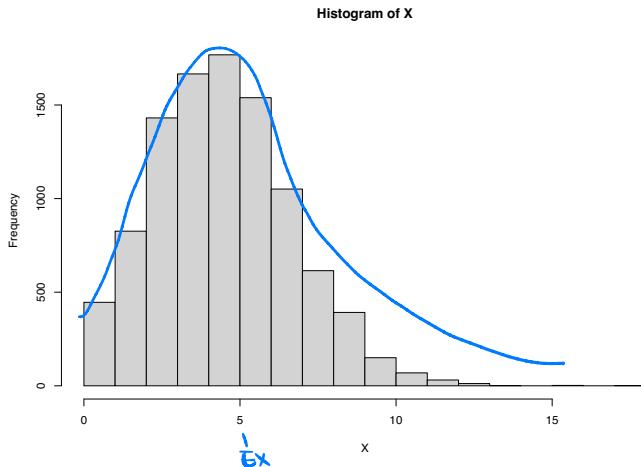
$$EX = \frac{1}{\lambda}$$

```
X <- rexp(n=10000, rate = 1/5) # meaning Lambda = 1/5
hist(X, breaks = 20)
```



Poisson random numbers

```
X <- rpois(n=10000, lambda = 5) # meaning Lambda=5  
hist(X, breaks = 20)
```



Exercise 2

Write a concise python code including histograms in this section. (p.17 ~ p.20)

"If I only had an hour to chop down a tree, I would spend the first 45 minutes sharpening my axe.
- A. Lincoln"