# STAT3006:TUTORIAL3

#### Sampling methods:

- Inverse Method.
- 2. Accept-Reject method.



### RANDOM SAMPLES

- In statistics, we use data to extract information, but how is the data produced?
- The most basic problem: how to get a sequence of (uniform) random numbers?
  - Pseudorandom number generator.
  - Why called pseudo? If you know the seed  $(X_0)$ , you know any following number in the sequence.
  - Examples:
    - linear congruential generator (often has poor performance):  $X_{n+1} = (aX_n + c) \mod m$ .
    - Mersenne Twister (most used).
  - Why called random? The sequence  $\{X_0, X_1, ..., X_n\}$  can pass numerous statistical tests for randomness (e.g. Diehard tests, Kolmogorov-Smirnov test).
  - Throughout the tutorial 3, we assume we can sample from uniform distribution on [0,1]



- We want to sample from F(x), where F(x) is a distribution function.
- Define the generalized inverse function of  $F, F^-(y)$ , as  $\inf\{x: F(x) \ge y\}$ .
- Inverse method:
  - Draw a sample U from unif[0,1];
  - Let X be  $F^{-}(U)$ .
- Some requirements:
  - F is a univariate function
  - $F^-$  has a closed form or  $F^-$  can be easily calculated numerically.



- A discrete case: drawing samples from Binomial(n, p)
- $p(x|n,p) = \binom{n}{x} p^x (1-p)^{n-x}$
- $F(x|n,p) = \sum_{t \le x} {n \choose t} p^t (1-p)^{n-t}$
- Inverse method:
  - *U* from unif[0,1]
  - If  $F(x|n, p) < U \le F(x + 1|n, p)$ , let X = x + 1.

```
    #samples from binomial distribution

• m <- 2000 # sample number
n <- 8 # trial number</li>

    p <- 0.4 # success probability</li>

x <- 0:n</p>
prob_mass <- factorial(n) / (factorial(n-x) * factorial(x)) * p^x * (1-p)^(n-x) #probability mass</pre>

    prob_cumsum <- cumsum(prob_mass) #cumulative sum of porbability masses</li>

u_vec <- runif(m) #uniform samples</li>

    x_vec <- NULL #binomial samples</li>

for(i in 1:m){
              x_{ec} <- c(x_{ec}, min(which(prob_cumsum >= u_vec[i])) - 1)
• }
hist(x_vec, breaks = seq(-0.5, 8.5, by = 1), freq = FALSE) #histogram
```

- A continuous case: drawing samples from a truncated normal distribution N(0,1)I(x>2.5)
- An intuitive approach is fist drawing samples from N(0,1) and then only selecting samples greater than 2.5.
- Is it valid? Yes, but not effective.
- Inverse method (what is  $F^-$ ):
  - *U* from unif[0,1]
  - Let  $X = F^{-}(U)$
  - $F^{-}(u) = \Phi^{-1}((1 \Phi(2.5))U + \Phi(2.5))$

- #samples from truncated normal distribution
- m <- 2000 #sample number</li>
- trun\_point <- 2.5</li>
- u\_vec <- runif(m) #uniform samples</li>
- const <- pnorm(2.5)</li>
- x\_vec <- qnorm((1-const)\*u\_vec + const )</pre>
- hist(x\_vec, breaks = 20, freq = FALSE) #histrogram



# ACCEPT-REJECT METHOD

- When F and the inverse of F is not tractable.
- We are interested in sampling from a pdf f(x).
- There exists a pdf g(x), s.t. f(x) < Mg(x).
- Sampling from g(x) is easy.
- Then we can apply accept-reject method:
  - Step1: draw Y  $\sim g(x)$ ,  $U \sim unif[0,1]$
  - Step2: if  $U < \frac{f(Y)}{Mg(Y)}$ , let X = Y.
  - The acceptance rate is 1/M. The smaller the M is, the more effective the accept-reject method is.



## ACCEPT-REJECT METHOD

- Example: we want to sample from a truncated normal distribution N(0,1)I(x > 2.5).
- How to determine the g(x)?
- What is f(x)/Mg(x)?
- g(x) is a shifted exponential pdf.

# ACCEPT-REJECT METHOD

```
    #samples from truncated normal distribution

    m <- 2000</li>

• i <- 0
• M <- \exp(-2.5^2/2)/(2.5* \operatorname{sqrt}(2*pi)*(1-pnorm(2.5)))
• ratio <- function(y){sqrt(2*pi)*exp(-y^2/2 + 2.5*(y-2.5) + 2.5^2/2)}
x vec <- NULL</li>
while(i <= m){</p>
              y < - \exp(1, 2.5) + 2.5
              u <- runif(1)
              if(u \le ratio(y))
                             x_{ec} <- c(x_{ec}, y)
                             i <- i+1
hist(x_vec, breaks = 20, freq = FALSE) #histrogram
```