Random Numbers and Law of Large Numbers

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Pseudo random number generators

- AbstractRandom class
 - java.util.Random class inside
 - Generate next random getNext()
- Transform method (変換法): Transform class
- Rejection method (棄却法): Rejection class

Sample Program

https://github.com/modeling-and-simulation-mc-saga/Random

Transform Method

ullet Probability density f(x) ($x\in [a,b)$) and probability distribution F

$$F(x) = \int_{a}^{x} f(z) dz \tag{1}$$

- Tranform method is available if the inverse of F(x) is obtained.
- Process
 - Generate a random number $r \in [0, 1)$.
 - $x = F^{-1}(r)$
 - $\{x\}$ distribute with f(x)

Example: Exponential distribution

$$f(x) = Ae^{-x}$$

$$0 \le x < 1$$
(2)

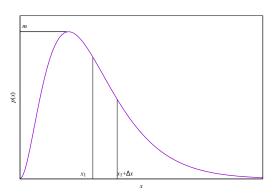
$$A = \frac{e}{e - 1} \tag{3}$$

$$F(x) = \int_0^x f(z) dz = A (1 - e^{-x})$$
 (4)

$$F^{-1}(r) = -\ln\left(1 - \frac{r}{4}\right) \tag{5}$$

Rejection Method

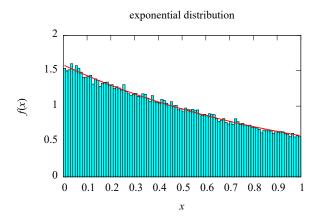
- Probability density f(x) defined in [a,b).
- Generate a random number pair $(x,y) \in [a,b) \times [0, \max f(x))$.
- Probability entering $[x_i, x_i + \Delta x]$ is proportional to $f(x)\Delta x$.



Classes for generating random numbers

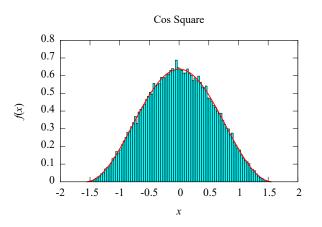
- randomNumbers package
 - AbstractRandom.java
 - Transform.java
 - Rejection.java
- Using java.util.Function.DoubleFunction
 - Define the inverse of F(x) for the Transform method.
 - Define f(x) for the Rejection method.
 - Using lambda expressions

Example of Transform method: exponential distribution



Example of Rejection Method: Square of Cosine

$$f(x) = \frac{2}{\pi}\cos^2(x), \quad x \in [-\frac{\pi}{2}, \frac{\pi}{2}]$$
 (6)



Probability and Law of Large Numbers (大数の法則)

- ullet What does it mean that the probability of getting 1 on a dice is 1/6?
- Consider the relative frequency of getting 1 on a dice.
 - ullet It approaches 1/6 with a large number of trials.
- Law of Large Numbers
- Let us take a closer look of this phenomenon.

Sample mean

- Consider a probabilistic variable X with the mean μ and deviation σ^2 .
- Sample mean with size n.

$$\bar{X} = \frac{1}{n} \sum_{k=0}^{n-1} X_k \tag{7}$$

- ullet Evaluate the population mean and deviation of $ar{X}$.
 - Evaluate the mean and deviation of \bar{X} with the probability of the population (母集団).
 - Equivalent to the mean of a large number of samples.

Population mean of sample means

The mean equals to the population mean μ .

$$E(\bar{X}) = \frac{1}{n} E\left(\sum X_k\right)$$

$$= \frac{1}{n} \sum E(X_k) = \frac{1}{n} n\mu$$

$$= \mu$$
(8)

Population deviation of sample means

The deviation reduces with n^{-1} .

$$V(\bar{X}) = E((\bar{X} - \mu)^{2}) = E\left(\frac{1}{n^{2}}\left(\sum_{k}(X_{k} - \mu)\right)^{2}\right)$$

$$= \frac{1}{n^{2}}E\left(\sum_{k}(X_{k} - \mu)^{2} + \sum_{i \neq j}(X_{i} - \mu)(X_{j} - \mu)\right)$$

$$= \frac{1}{n^{2}}E\left(\sum_{k}(X_{k} - \mu)^{2}\right) + \frac{1}{n^{2}}E\left(\sum_{i \neq j}(X_{i} - \mu)(X_{j} - \mu)\right)$$

$$= \frac{1}{n^{2}}n\sigma^{2} = \frac{\sigma^{2}}{n}$$
(9)

Confirm the law of large numbers by simulations

- Generate samples of size n.
- Instead evaluating the mean using the population distribution
 - ullet Generate a large number m of samples with the same size.
 - Evaluate the mean and deviation for samples
- ullet Changing n and observe n dependence.

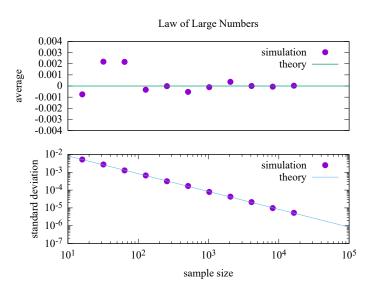
Example: uniform

$$f(x) = \begin{cases} 1 & -\frac{1}{2} \le x < \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$
 (10)

$$\langle x \rangle = \int_{-1/2}^{1/2} x f(x) dx = \int_{-1/2}^{1/2} x dx = \left[\frac{1}{2} x^2 \right]_{-1/2}^{1/2} = 0$$
 (11)

$$\langle x^2 \rangle = \int_{-1/2}^{1/2} x^2 f(x) dx = \int_{-1/2}^{1/2} x^2 dx = \left[\frac{1}{3} x^3 \right]_{-1/2}^{1/2} = \frac{1}{12}$$
 (12)

$$\sigma^2 = \langle x^2 \rangle - \langle x \rangle = \frac{1}{12} \tag{13}$$



m = 1000 for each sample size.

Pseudo Random Numbers (疑似乱数)

- Generate random numbers using a computer.
- Need some kind of algorithms.
 - Sequences are deterministic.
- In other word, you can generate an identical sequence as you need.

Linear Congruential Method (線形合同法)

- Recursion relation: $v_n = (av_{n-1} + c) \mod m$
 - ullet Property strongly depends on parameters a, c, m
 - Good parameter sets are known empirically.
- Unsigned integers are available in C/C++
 - No need to control overflow
- Languages, such as Java and Fortran, does not have unsigned integers.
 - Need some tips to suppress overflow

• Parameter example for 32 bit unsigned integers

$$m = 2416, c = 374441, m = 1771874$$

• Parameter example for 32 bit signed integers

$$m = 9301, c = 49297, m = 233280$$

Schrage's method

For 32 bit signed integers

$$\bullet$$
 Set $m=2^{31}-1$
$$a=17807,\ c=0,\ m=2^{31}-1$$

$$q=\lfloor m/a\rfloor$$

$$r=m\ \mathrm{mod}\ a$$

$$m=aq+r$$

• Need the condition r < q

$$av_{n+1} \bmod m = \begin{cases} a (v_n \bmod q) - \lfloor v_n/q \rfloor r & \text{if not negative} \\ a (v_n \bmod q) - \lfloor v_n/q \rfloor r + m & \text{otherwise} \end{cases}$$
(14)

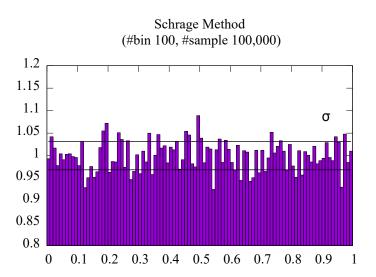
- Because, consider $v_n = xq + y$
 - LHS: ay xr
 - And

$$av_{n+1} = a(xq + y) = xaq + ay$$

= $x(m-r) + ay = xm + ay - xr$ (15)

$$av_{n+1} \bmod m = ay - xr \tag{16}$$

Example of Schrage's method



Difficulties in LCM

- The next value of a value a is determined.
- The period is limited by m.
 - ullet Some kinds of simulation need a larger number of random numbers than m.
- Multidimensional sparse crystal (多次元粗結晶)
 - Consider consecutive n random numbers as a point in n dimensional space.
 - Those have crystal structure with some parameter set.