

# RANDOM NUMBERS

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# UNIFORM RANDOM NUMBERS

Task: producing  $X_1, \dots, X_n \sim U[0, 1]$ . Start with some initial number  $X_0$  and use a transformation  $D$  such that  $X_{i+1} = D(X_i)$ .

Example: linear congruential generators have a period  $M$

$$x_{i+1} = a \cdot x_i + b(\text{modulo } M).$$

Usually  $M$  is very large ( $2^{59}$ ).

Remember: any sequence of random numbers will eventually repeat itself. You need to set the seed of your random number generator to obtain a different sequence of numbers. Getting the same sequence of numbers is actually good for replicating results and debugging.

# THE INVERSE TRANSFORM

The generalized inverse of a non-decreasing function  $F$  is:

$$F^-(u) = \inf\{x : F(x) \geq u\}.$$

## LEMMA

*If  $U \sim U[0, 1]$ , then the random variable  $F^-(U)$  has distribution  $F$ .*

## PROOF.

$$P(F^-(U) \leq x) = P(U \leq F(x)) = F(x).$$



## EXAMPLE

If  $X \sim \text{Exp}(1)$  we have  $F(x) = 1 - \exp(-x)$ . Thus  $F^-(u) = -\log(1 - u)$ . Therefore, if you simulate  $U \sim U[0, 1]$ , the r.v.  $X = -\log(U) \sim \text{Exp}(1)$ .

# EXPONENTIAL(1) LEADS TO OTHER DISTRIBUTIONS

Let  $X_1, X_2, \dots$  be i.i.d.  $Exp(1)$ . Then:

$$2 \sum_{j=1}^k X_j \sim \chi_{2k}^2,$$

$$\beta \sum_{j=1}^a X_j \sim \text{Gamma}(a, \beta),$$

$$\frac{\sum_{j=1}^a X_j}{\sum_{j=1}^{a+b} X_j} \sim \text{Beta}(a, b),$$

and so on.

# GENERATING RANDOM NORMALS $N(0,1)$

## THE BOX-MULLER ALGORITHM

- Generate  $U_1, U_2$  i.i.d.  $U[0,1]$ .
- Define

$$\begin{aligned}x_1 &= \sqrt{-2 \log(u_1)} \cos(2\pi u_2), \\x_2 &= \sqrt{-2 \log(u_2)} \sin(2\pi u_2).\end{aligned}$$

- Take  $x_1$  and  $x_2$  as two independent draws from  $N(0,1)$ .

# GENERATING RANDOM POISSON $Pois(\lambda)$

If  $N \sim Pois(\lambda)$  and  $X_1, X_2, \dots$  i.i.d.  $Exp(\lambda)$ , then

$$P(N = k) = P(X_1 + \dots + X_k \leq 1 < X_1 + \dots + X_k + X_{k+1}).$$

You simulate i.i.d.  $Exp(\lambda)$  random variables until their sum exceeds 1.  
The number of variables you simulated is a random draw from  $Pois(\lambda)$ .

# GENERATING DISCRETE RANDOM VARIABLES

Let  $X$  be a r.v. that takes values  $\{1, 2, \dots, K\}$ . Calculate the cumulative probabilities:

$$\begin{aligned}p_0 &= 0, \\p_1 &= P(X \leq 1) = P(X = 1), \\p_2 &= P(X \leq 2) = p_1 + P(X = 2), \\&\vdots \\p_K &= P(X \leq K) = p_{K-1} + P(X = K) = 1.\end{aligned}$$

Generate  $U \sim U[0, 1]$ . Take  $X = k$  if  $p_{k-1} < U \leq p_k$ .

# GENERATING MULTIVARIATE NORMAL RANDOM VARIABLES

Need to generate from the  $p$ -dimensional normal  $N_p(\mu, \Sigma)$ . Obtain the Cholesky factorization of  $\Sigma$

$$\Sigma = (\Sigma^{1/2})^T \Sigma^{1/2},$$

where  $\Sigma^{1/2}$  is an upper triangular matrix with positive elements (use BLAS!). Simulate  $p$  independent  $N(0, 1)$  r.v.  $X_1, \dots, X_p$ . Then

$$\mu + \Sigma^{1/2} \begin{bmatrix} X_1 \\ \vdots \\ X_p \end{bmatrix} \sim N_p(\mu, \Sigma).$$



`http://www.gnu.org/software/gsl/`