

MAT 653: Statistical Simulation

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Random Variable Generation

Reading Parts:

1. Chapter 2 from Introducing Monte Carlo Methods with R, by Christian. P. Robert and George Casella.
2. Section 1.4.1, 1.4.2, 1.4.3, 1.4.4 from An Introduction to Statistical Computing: A Simulation-based Approach, by Jochen Voss.

Definition

A random variable U has a uniform distribution $U \sim U(0, 1)$ if $P(U \in (a, b)) = b - a$ for $0 < a < b < 1$.

Definition

A random variable has PDF (probability density function) $f(x)$ if $P(X \in A) = \int_A f(x)dx$.

- **PDF** $f(x)$:

1. $f(x) \geq 0$
2. $\int_{\mathbb{R}} f(x)dx = 1$

The support of X is $\{x : f(x) > 0\}$ denoted by \mathcal{X} .

- **CDF**: the CDF of X (on \mathbb{R}) is given by $F(a) = P(X \leq a)$, for all $a \in \mathbb{R}$.

1. $\lim_{x \rightarrow -\infty} F(a) = 0$
2. $\lim_{x \rightarrow \infty} F(a) = 1$
3. $F(a)$ is nondecreasing function.
4. $\lim_{h_n \rightarrow 0^+} F(a + h_n) = F(a)$. i.e., $F(a)$ is right continuous.

Fact: Any non-negative function \tilde{f} that is integrable on its support can be used to construct a PDF by normalization.

Examples:

1. $\tilde{f}(x) = e^{-\frac{x^2}{2}}$, $\mathcal{X} = (-\infty, \infty)$

$$f(x) = \frac{\tilde{f}(x)}{\int_{-\infty}^{\infty} \tilde{f}(x)dx} \text{ (normalization)} = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \text{ where } e^{-\frac{x^2}{2}} \text{ is the kernel of } N(0, 1).$$

$$f(x) \propto e^{-\frac{x^2}{2}}.$$

$$2. \tilde{f}(x) = x^{\alpha-1} e^{-\beta x}, x > 0, \alpha, \beta > 0$$

$$f(x) = \frac{\tilde{f}(x)}{\int_0^\infty x^{\alpha-1} e^{-\beta x} dx} = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} \quad (x > 0) \sim \text{Gamma}(\alpha, \beta)$$

$$f(x) \propto x^{\alpha-1} e^{-\beta x}$$

Probability integral transform

(works only for \mathbb{R})

Suppose X has a CDF F , i.e. $F(x) = P(X \leq x)$. If F is invertible and $Y = F^{-1}(U)$. $U \sim U(0, 1)$.

Then $Y \sim F$.

Proof:

More generally, define the inverse of F as $F^{-1}(u) = \inf \{x \in \mathbb{R} : F(x) \geq u\}$.

Fact1: For every $0 < p < 1$, $F \circ F^{-1}(p) \geq p$. The equality holds iff (if and only if) p in the range of F .

Fact2: For every $0 < p < 1$ and $x_0 \in \mathbb{R}$, $p \leq F(x_0)$ iff $F^{-1}(p) \leq x_0$.

Proof:

Example 2.1 from R.C.:

$X \sim \text{exp}(1)$, the CDF of $\text{exp}(1)$ is $F(x) = 1 - e^{-x}$.

Then $F^{-1}(u) = -\log(1 - u)$.

1. Generate $u \sim U(0, 1)$.

2. Evaluate $F^{-1}(u)$, gives one sample.