

Generating continuous Random variables

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The Inverse transform algorithm

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- *Proposition:* Let $U \sim U(0, 1)$. For any continuous CDF F , the random variable X defined by
$$X = F^{-1}(U)$$
has distribution F .
- *proof* in class.
- So long as we can derive the explicit form for F^{-1} , we can use this result to generate from a continuous distribution with CDF F :
 - step 1: generate a random number U .
 - step 2: set $X = F^{-1}(U)$ and you are done.

proof

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$X = F^{-1}(U)$ has a cdf of F .

Look at cdf

$$\begin{aligned} P(X \leq x) = F_x(X) &= P(F(F^{-1}(U)) \leq F(x)) \\ &= P(U \leq F(x)) \\ &= F(x) \end{aligned}$$

therefore $F(x)$ is the cdf of $F^{-1}(U)$.

Some examples

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- Let X come from a distribution with CDF

$$F(x) = x^n, 0 < x < 1.$$

We can generate from this distribution by setting $X = U^{1/n}$ after generating U .

- *Exponential* Suppose X comes from an $Exp(\lambda)$ with PDF

$$f(x) = \lambda e^{-\lambda x}, x > 0.$$

We can generate from this distribution by setting $X = -\frac{1}{\lambda} \log U$ where U is the generated random number.

- *Weibull* A random variable X is said to have a Weibull(α, β) distribution if its CDF has the form

$$F(x) = 1 - \exp(-\alpha x^\beta), x > 0.$$

Describe a method to generate from this distribution.

Generating a Gamma random variable

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- For a $\text{Gamma}(n, \lambda)$ random variable, note that because we cannot write an explicit form for the expression of the F^{-1} , it is difficult to directly apply inverse transform method.
- However, recall that the sum of independent Exponentials leads us to a Gamma distribution.
- We can generate from a Gamma distribution by generating n random numbers U_1, U_2, \dots, U_n and then setting
$$X = -\frac{1}{\lambda} \log U_1 - \dots - \frac{1}{\lambda} \log U_n$$
$$= -\frac{1}{\lambda} \log(U_1 \cdots U_n).$$
- This works provided n is a positive integer.

The rejection method

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- Suppose we wish to generate X from a distribution with PDF $f(x)$.
- Assume that we are able to generate Y from a distribution with PDF $g(y)$ and that there is a constant c such that $\frac{f(y)}{g(y)} \leq c$, for all y .
- According to the rejection method, we can generate X using the following steps:
 - step 1: generate Y from distribution with density g .
 - step 2: generate a random number U .
 - step 3: if $U \leq \frac{f(Y)}{cg(Y)}$, set $X = Y$.
 - step 4: else return to step 1.

Important results

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$$X \sim U(a, b), \quad a < b$$

use inverse transform method

$$F(x) = \int_a^x \frac{1}{b-a} dz = \frac{x-a}{b-a}$$

■ Generate a random number U

■ $\frac{x-a}{b-a} = U \implies X = a + U(b-a).$

$$\text{Examples: } \begin{aligned} X \sim U(0, 2\pi) &\implies X = 2\pi U \\ X \sim U(-1, 1) &\implies X = 2U - 1. \end{aligned}$$

–continued

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Theorem:

- The random variable X generated by the rejection method has density f .
- The number of iterations required for the rejection algorithm has a geometric distribution with mean c .

proof: to be discussed in class.

Some examples

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- *Example5d*: Use the rejection method to generate from $f(x) = 20x(1-x)^3, 0 < x < 1$.
- *Example5e*: Use the rejection method to generate from a `Gamma(shape para=3/2, scale = 1)` density with $f(x) = Kx^{1/2}e^{-x}, x > 0$.
where $K = 1/\Gamma(3/2) = 2/\sqrt{\pi}$.