#### 1: Inverse method for Poissson Distribution (25%)

For discrete Poisson Distribution ( $\lambda = 5$ ),

the p.m.f is  $P(x|\lambda) = e^{-\lambda} \frac{\lambda^x}{x!}$  and the c.d.f is  $F(x|\lambda) = \sum_{t \le x} e^{-\lambda} \frac{\lambda^t}{t!}$ .

**Algorithm:** Inverse method for the Poisson Distribution:

To generate  $X \sim F(x)$ :

STEP 1: Generate  $U \sim unif[0,1]$ ;

STEP 2: Transform  $X = F^{-}(U)$ : if  $F(x|\lambda) < U \le F(x+1|\lambda)$ , let X = x+1.

Plot:

# Histogram of x\_vec

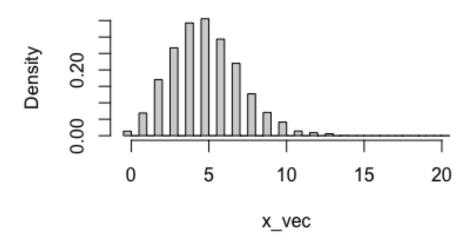


Figure 1: Histogram of 5000 samples

# 2: Accept-Reject method for truncated Gamma Distribution (25%)

For 
$$X \sim Gamma(\frac{1}{2}, 1)I(x \ge 5)$$
,  $f(x) = \frac{x^{-\frac{1}{2}}e^{-x}I(x \ge 5)}{\int_{5}^{+\infty} y^{-\frac{1}{2}}e^{-y}dy}$ .

We can define a shifted exponential distribution  $g(x) = e^{-(x-5)}I(x \ge 5)$  and want to find a constant M such that f(x) < Mg(x) for any x.

Then 
$$M = \sup \frac{f(x)}{g(x)} = \sup \frac{\frac{x^{-\frac{1}{2}}e^{-x}I(x \ge 5)}{\int_{5}^{+\infty}y^{-\frac{1}{2}}e^{-y}dy}}{e^{-(x-5)}I(x \ge 5)} = \frac{5^{-\frac{1}{2}}e^{-5}}{\int_{5}^{+\infty}y^{-\frac{1}{2}}e^{-y}dy}.$$

Algorithm: Accept-Reject method for truncated Gamma Distribution:

To generate  $X \sim F(x) = \text{c.d.f of } f(x)$ :

STEP 1: Generate  $Y \sim g(y)$ ;

STEP 2: Generate  $U \sim unif[0,1]$ ;

STEP 3: Accept X = Y if  $U \leq \frac{f(Y)}{Ma(Y)}$ .

#### Proof:

From the choice of constant M, we can know that  $Mg(x) \ge f(x)$ . The goal of this method is to generate  $X \sim F(x) = \text{c.d.f}$  of f(x).

For the generating algorithm:

$$P(X \le x) = P(Y \le x | Y \text{ is accepted})$$

$$= P(Y \le x | U \le \frac{f(Y)}{Mg(Y)})$$

$$= \frac{P(Y \le x, U \le \frac{f(Y)}{Mg(Y)})}{P(U \le \frac{f(Y)}{Mg(Y)})}$$

$$= \frac{\int_{-\infty}^{x} g(y) \int_{0}^{\frac{f(y)}{Mg(y)}} 1 du dy}{\int_{-\infty}^{+\infty} g(y) \int_{0}^{\frac{f(y)}{Mg(y)}} 1 du dy}$$

$$= \frac{\int_{-\infty}^{x} g(y) \frac{f(y)}{Mg(y)} dy}{\int_{-\infty}^{+\infty} g(y) \frac{f(y)}{Mg(y)} dy}$$

$$= \frac{\int_{-\infty}^{x} f(y) dy}{\int_{-\infty}^{+\infty} f(y) dy}$$

$$= \frac{\int_{-\infty}^{x} f(y) dy}{1}$$

$$= F(x)$$

Therefore, this AR method works.

#### Comparison:

Theoretical acceptance probability:

$$\begin{split} P(U \leq \frac{f(Y)}{Mg(Y)}) &= \int_{-\infty}^{+\infty} g(y) \int_{0}^{\frac{f(y)}{Mg(y)}} 1 du dy \\ &= \frac{1}{M} \int_{-\infty}^{+\infty} f(y) dy \\ &= \frac{1}{M} \end{split}$$

After computation, this acceptance probability is 0.184157.

The actual acceptance rate is 0.1854, which is a little bit higher than the theoretical value.

### 3: Importance Sampling for Estimation (25%)

(1) Using 5000 samples from Q2 (l = length of samples obtained in Q2), the Monte Carlo estimate is

$$\int_{5}^{+\infty} \cos(x) x^{-\frac{1}{2}} e^{-x} dx = \int_{-\infty}^{+\infty} \cos(x) \frac{x^{-\frac{1}{2}} e^{-x} I(x \ge 5)}{\int_{5}^{+\infty} y^{-\frac{1}{2}} e^{-y} dy} dx \times \int_{5}^{+\infty} y^{-\frac{1}{2}} e^{-y} dy$$

$$= \frac{\int_{5}^{+\infty} y^{-\frac{1}{2}} e^{-y} dy}{l} \sum_{i=1}^{l} \cos(x_i)$$

$$= 0.001708$$

(2) Using the same notations in Q2, we define h(x) = cos(x).

$$\int_{5}^{+\infty} \cos(x) x^{-\frac{1}{2}} e^{-x} dx = \int_{-\infty}^{+\infty} \cos(x) \frac{x^{-\frac{1}{2}} e^{-x} I(x \ge 5)}{\int_{5}^{+\infty} y^{-\frac{1}{2}} e^{-y} dy} dx \times \int_{5}^{+\infty} y^{-\frac{1}{2}} e^{-y} dy$$
$$= \int_{5}^{+\infty} y^{-\frac{1}{2}} e^{-y} dy \int_{-\infty}^{+\infty} \frac{h(x) f(x)}{g(x)} dx$$

Note that  $\frac{f(x)}{g(x)} \le M < \infty$ , and  $E_g h^2(x) = \int_{-\infty}^{+\infty} g(x) h^2(x) dx \le \int_{-\infty}^{+\infty} g(x) dx = 1 < \infty$ , we can use the importance sampling as follows:

## Algorithm:

STEP 1: Generate n = 5000 samples from g(x) by Accept-Reject Method;

STEP 2: Compute the Monte Carlo estimate:

$$m(x) = \frac{\sum_{i=1}^{n} \frac{h(x_i)f(x_i)}{g(x_i)}}{n} = \frac{e^{-5}}{n \int_{5}^{+\infty} y^{-\frac{1}{2}} e^{-y} dy} \sum_{i=1}^{n} \cos(x_i) x_i^{-\frac{1}{2}}$$

# 4: Stratified Sampling (25%)

- **(1)**
- **(2)**
- (3)