Notation: here is a brief summary of the notation used in this worksheet.

- p(X = x) is equal to the probability density function;
- $\bullet$  Capital letters such as X stand for the random variable.

## Exercise 1 (Inversion and Rejection)

1. Let  $F_X(x) = \mathbb{P}(X \leq x)$  and  $U \sim Unif[0, 1]$ :

$$F_X(x) = 1 - e^{-\lambda(X-a)} \mathbb{I}_{\{X \ge a\}} = U$$
$$-\ln(1-U) = \lambda(x-a)$$
$$F_X^{-1}(U) = a - \frac{\ln(1-U)}{\lambda}$$

To simulate X from U, just simulate value from U and substitute in the formula above.

2. Let  $X = Y \mid a \leq Y \leq b$ . First, let's show that  $X = F_Y^{-1}(F_Y(a)(1-U) + F_Y(b)U)$ :

$$\mathbb{P}(X \le x) = \mathbb{P}(F_Y^{-1}(F_Y(a)(1-U) + F_Y(b)U) \le x) = \mathbb{P}(F_Y^{-1}(F_Y(a) + U[F_Y(b) - F_X(a)]) \le x)$$

$$= \mathbb{P}(F_Y(a) + U[F_Y(b) - F_X(a)] \le F_Y(x)) = \mathbb{P}\left(U \le \frac{F_Y(x) - F_Y(a)}{F_Y(b) - F_Y(a)}\right) = \frac{F_Y(x) - F_Y(a)}{F_Y(b) - F_Y(a)}$$

Note that since  $x \in [a, b]$ :

$$\mathbb{P}(Y \le x \mid a \le Y \le b) = \frac{\mathbb{P}(Y \le x, a \le Y \le b)}{\mathbb{P}(a \le Y \le b)} = \frac{\mathbb{P}(a \le Y \le x)}{F_Y(b) - F_Y(a)} = \frac{F_Y(x) - F_Y(a)}{F_Y(b) - F_Y(a)} = \mathbb{P}(X \le x)$$

Now that we proved the above relation, to simulate an exponential conditioned on  $\geq a$ , we first generate  $U \sim Unif[0,1]$ , then, for  $Y \sim Expo(\lambda)$ :

$$F_Y(y) = 1 - e^{\lambda y} : F_Y^{-1}(U) = \frac{-\ln(1 - U)}{\lambda}$$

$$X = \frac{-\ln(1 - (1 - U)F_Y(a) + U)}{\lambda} = \frac{-\ln(e^{-\lambda a} + U \cdot e^{-\lambda a})}{\lambda} = a - \frac{\ln(1 - U)}{\lambda}$$

The formula yields the same solution as the one obtained using inversion.

## Worksheet 1

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3. Let 
$$q \sim Expo(\lambda)$$
, and  $\pi(x) = \lambda e^{-\lambda(x-a)} \mathbb{I}_{x \geq a}$ :

Note that 
$$M = \max_x \pi(x)/q(x) = e^{\lambda a}$$
, since  $\pi(x)/q(x) = \frac{\lambda e^{-\lambda(x-a)}}{\lambda e^{\lambda(x)}} = e^{\lambda a}$ 

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In the rejection method, we sample  $x_i \sim q$ ,  $u \sim Unif[0,1]$ , then we accept a sample  $x_i$  if  $u_i \leq \frac{\pi(x_i)}{Mq(x_i)}$ .

Hence,

- If  $x \le a \implies \pi(x) = 0 \implies u \le 0$ :  $x_i$  is rejected;
- If  $x > a \implies \pi(x) = 1 \implies u \le 1 :: x_i$  is accepted;

Which is the same procedure described in the question, implying that it is equal to the rejection algorithm.

Finally, the expected number of trials is equal to  $M = e^{\lambda a}$ . Therefore, for  $a \gg 1/\lambda$ , the expected number of trials becomes very large (greater computational cost), while this problem doesn't happen with inversion, since every sample is used.

## Exercise 2 (Rejection)

1. Let A be the event where the value is accepted at some point, while  $A_b$  is accepted at step (b) and  $A_c$  is accepted at step (c):

In step (b) we have:

$$\mathbb{P}(x \in A_b) = \frac{h(x)}{M\tilde{q}(x)}$$

In step (c) we have:

$$\mathbb{P}(x \in A_c) = \frac{\tilde{\pi}(x) - h(x)}{M\tilde{q}(x) - h(x)}$$

Since step (b) is independent of (c),

$$\mathbb{P}(x \in A) = \mathbb{P}(x \in A_b \cup x \in A_c) =$$

$$= \mathbb{P}(x \in A_b) + \mathbb{P}(x \in A_b) - \mathbb{P}(x \in A_b \cap x \in A_c) =$$

$$= \frac{h(x)}{M\tilde{q}(x)} + \frac{\tilde{\pi}(x) - h(x)}{M\tilde{q}(x) - h(x)} + \frac{h(x)}{M\tilde{q}(x)} \cdot \frac{\tilde{\pi}(x) - h(x)}{M\tilde{q}(x) - h(x)} = \frac{\tilde{\pi}(x)}{M\tilde{q}(x)}$$

2. Let B be an arbitrary event.

$$\mathbb{P}(X \in B \mid X \in A) = \mathbb{P}(X \in B \cap X \in A)/\mathbb{P}(X \in A) ::$$

$$\mathbb{P}(X \in B \cap X \in A) = \int_{\chi} \int_{0}^{1} \mathbb{I}_{B}(x) \mathbb{I}\left(u \leq \frac{\tilde{\pi}(x)}{M\tilde{q}(x)}\right) q(x) du dx$$

$$\mathbb{P}(X \in B \cap X \in A) = \int_{B} \frac{\tilde{\pi}(x)}{M\tilde{q}(x)} \tilde{q}(x) \cdot Z_{q}^{-1} dx$$

$$\mathbb{P}(X \in B \cap X \in A) = \frac{\pi(B) \cdot Z_{\pi}}{M \cdot Z_{q}}$$

Finally,

$$\mathbb{P}(X \in A) = \frac{Z_{\pi}}{M \cdot Z_q} : \mathbb{P}(X \in B \mid X \in A) = \frac{\pi(B)Z_{\pi}}{MZ_q} \cdot \frac{MZ_q}{Z_{\pi}} = \pi(B)$$

3. We want to show that:

$$\mathbb{P}(\text{Step (c) is necessery}) = 1 - \frac{\int_{\chi} h(x) dx}{MZ_q}$$

First, note that  $\mathbb{P}(\text{Step }(c) \text{ is necessery}) = \mathbb{P}(X \text{ not accepted in step}(b))$ , hence:

$$\mathbb{P}(X \text{ not accepted in step(b)}) = 1 - \mathbb{P}(X \in A_b) = 1 - \mathbb{P}\left(U \le \frac{h(X)}{M\tilde{q}(X)}\right)$$

$$\mathbb{P}(X \in A_b) = \mathbb{P}(X \in X \cap X \in A_b) =$$

$$\int_{\chi} \int_{0}^{1} \mathbb{I}_{\chi}(x) \mathbb{I}\left(u \le \frac{h(x)}{M\tilde{q}(x)}\right) du dx =$$

$$\int_{\chi} \frac{h(x)\tilde{q}(x)}{M\tilde{q}(x)Z_q} dx = \frac{\int_{\chi} h(x) dx}{MZ_q} :$$

$$\mathbb{P}(\text{Step(c) is necessery}) = 1 - \frac{\int_{\chi} h(x) dx}{MZ_q}$$

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4. We want to calculate the probability of not having to evaluate  $\tilde{pi}(x)$ , which is equal to the probability of accepting the sample in step (b).

First, we know that:

$$\mathbb{P}(X \text{ is accepted in step(b)}) = \frac{\int_{\chi} h(x)dx}{MZ_q}$$

.

Note that  $h(x) \ge 0$ , therefore  $1 - x^2/2 \ge 0$ .  $-\sqrt{2} \le x \le \sqrt{2}$ . Hence,  $h(x) = 0, \forall x \notin [-\sqrt{2}, \sqrt{2}]$ .

$$\int_{-\sqrt{2}}^{\sqrt{2}} h(x)dx = \int_{-\sqrt{2}}^{\sqrt{2}} 1 - \frac{x^2}{2} dx = \frac{4\sqrt{2}}{3}$$

$$\int_{-\infty}^{\infty} e^{-|x|} dx = Z_q = 2 \int_0^{\infty} e^{-x} dx = 2(-[e^{-\infty} - e^0]) = 2$$

$$M = \sup_{x \in \mathbb{R}} \frac{\tilde{\pi}(x)}{\tilde{q}(x)} = \sup_{x \in \mathbb{R}} \frac{e^{-x^2/2}}{e^{-|x|}} = \sup_{x \in \mathbb{R}} e^{-x^2/2 + |x|}$$

For  $x \ge 0$ , we have  $\frac{d}{dx}(x - x^2/2) = 1 - x = 0 \implies x = 1$ .

For x < 0, we have  $\frac{d}{dx}(-x - x^2/2) = -1 - x = 0 \implies x = -1$ . Hence,  $M = \sqrt{e}$ .

Finally, we have:

$$\mathbb{P}(\mathbf{X} \text{ is accepted in step(b)}) = \frac{\int_{\chi} h(x) dx}{MZ_q} = \frac{4\sqrt{2}}{3 \cdot 2 \cdot \sqrt{e}} = \frac{2\sqrt{2}e^{-1}}{3}$$

It can be beneficial to use this algorithm instead of the standard rejection sampling procedure because this algorithm as more computationally efficient.

## Exercise 3 (Transformation)

1. Let  $V = arctan(U_2/U_1)$  and  $Y = U_1^2 + U_2^2 \le 1$ . We want to show that:

$$p_{Y,V}(y,\theta) = \mathbb{I}_{[0,1]}(y) \frac{\mathbb{I}_{[0,2\pi]}(\theta)}{2\pi}$$

First, note that  $p_{Y,V}(y,\theta) = p_{U_1,U_2}(u_1,u_2) \begin{vmatrix} \frac{\partial u_1}{\partial y} & \frac{\partial u_1}{\partial \theta} \\ \frac{\partial u_2}{\partial y} & \frac{\partial u_2}{\partial \theta} \end{vmatrix}$ 

$$\begin{vmatrix} \frac{\partial u_1}{\partial y} & \frac{\partial u_1}{\partial \theta} \\ \frac{\partial u_2}{\partial y} & \frac{\partial u_2}{\partial \theta} \end{vmatrix} = \begin{vmatrix} \frac{\partial y}{\partial u_1} & \frac{\partial \theta}{\partial u_1} \\ \frac{\partial y}{\partial u_2} & \frac{\partial \theta}{\partial u_2} \end{vmatrix}^{-1} = \begin{vmatrix} 2u_1 & \frac{1}{1 + (u_1/u_2)^2} \cdot \frac{-1u_2}{u_1^2} \\ 2u_2 & \frac{1}{1 + (u_1/u_2)^2} \cdot \frac{1}{u_1^2} \end{vmatrix}^{-1} = \frac{2}{1 + (u_2/u_1)^2} + \frac{2(u_2/u_1)^2}{1 + (u_2/u_1)^2} = \frac{2}{1 + \theta^2} + \frac{2(1 + \theta^2)}{1 + \theta^2} = 2$$

$$p_{Y,V}(y,\theta) = p_{U_1,U_2}(u_1,u_2) \cdot \frac{1}{2} = p_{U_1,U_2}(\sqrt{y}cos(\theta), \sqrt{y}sin(\theta)) \cdot \frac{1}{2}$$

Note that  $p_{U_1,U_2}(\sqrt{y}cos(\theta),\sqrt{y}sin(\theta))$  is a uniform distribution over a circle of radius  $Y \leq 1$ , therefore, it's normalizing constant is equal to pi, which is the area of the circumference of radius 1. With that we can write:

$$p_{Y,V}(y,\theta) = \mathbb{I}_{[0,1]}(y) \frac{\mathbb{I}_{[0,2\pi]}(\theta)}{2\pi}$$