

## Exercises 4.2–4.5: Inverse Transform & Rejection Sampling

- **Inverse transform sampling:** derive  $F$  and compute  $F^{-1}$ , then set  $X = F^{-1}(U)$  with  $U \sim \text{Unif}(0, 1)$ .
- **Rejection sampling:** choose easy proposal  $q$ , find  $c$  such that  $\pi(x) \leq cq(x)$ , accept with probability  $\pi(y)/(cq(y))$ .

## Exercise 4.2: normalise and invert

**Given:**

$$\pi(x) = ax^2, \quad x \in [0, 1].$$

Normalisation:

$$1 = \int_0^1 ax^2 dx = a \cdot \frac{1}{3} \quad \Rightarrow \quad a = 3.$$

CDF:

$$F(x) = \begin{cases} 0, & x < 0, \\ x^3, & 0 \leq x \leq 1, \\ 1, & x > 1. \end{cases} \quad F^{-1}(u) = u^{1/3}.$$

## Exercise 4.2 (R code, part 1/3): comments + simulate

```
#Integrating over [0, 1] and setting the result equal to 1, gives a = 3.  
#The CDF is 0 for x < 0, x^3 for 0 <= x <= 1 and 1 for x > 1.  
#The inverse function for the inverse transform method is u^1/3.  
  
#Generate random uniform samples  
u <- runif(10000)  
  
#Generate samples from required distribution  
x <- u^(1/3)
```

## Exercise 4.2 (R code, part 2/3): histogram + density overlay

```
#Plot results  
X <- seq(0, 1, 0.01)  
hist(x, freq = FALSE, main = "", breaks = 20)  
lines(X, 3*X^2)
```

## Exercise 4.2: what you should see

Histogram should match the curve  $\pi(x) = 3x^2$  on  $[0, 1]$ :

- few samples near 0,
- many samples near 1,
- density increases like  $x^2$ .

## Exercise 4.3: inverse transform

**Given:**

$$\pi(x) = \frac{1}{\theta} x^{\frac{1-\theta}{\theta}} = \frac{1}{\theta} x^{\frac{1}{\theta}-1}, \quad x \in [0, 1].$$

CDF:

$$F(x) = x^{1/\theta} \quad \Rightarrow \quad F^{-1}(u) = u^\theta.$$

We generate 10,000 samples for  $\theta \in \{1, 5, 10\}$ .

## Exercise 4.3 (R code, part 1/4): comments + function

```
#The inverse function is  $F^{-1}(x) = x^{\theta}$  for  $x \in [0, 1]$ .  
  
# Inverse Transform Function -----  
  
#This function take in a vector of  $U[0, 1]$  numbers and a value for theta  
#and returns samples from the distributions  $1/\theta x^{\{(1-\theta)/\theta\}}$ .  
  
inverse.transform.function <- function(u, theta){  
  
  #Transform using inverse distribution function  
  x <- utheta  
  
  return(x)  
}
```

## Exercise 4.3 (R code, part 2/4): generate samples

```
#Generate random uniform samples  
u <- runif(10000)  
  
#Generate samples from required distribution  
x1 <- inverse.transform.function(u, 1)  
x5 <- inverse.transform.function(u, 5)  
x10 <- inverse.transform.function(u, 10)
```

## Exercise 4.3 (R code, part 3/4): $\theta = 1$ and $\theta = 5$

```
#Plot results  
X <- seq(0, 1, 0.01)  
hist(x1, freq = FALSE, main = "", breaks = 20)  
lines(X, rep(1, length(X)))  
  
#Plot results  
X <- seq(0, 1, 0.01)  
hist(x5, freq = FALSE, main = "", breaks = 100)  
lines(X, 0.2*X^-0.8)
```

## Exercise 4.3 (R code, part 4/4): $\theta = 10$

```
#Plot results  
X <- seq(0, 1, 0.01)  
hist(x10, freq = FALSE, main = "", breaks = 100)  
lines(X, 0.1*X^-0.9)
```

## Exercise 4.4: rejection sampling on $[0, 1]$

**Target:**  $\pi(x) = 3x^2$  on  $[0, 1]$ .

Use proposal  $q(x) = 1$  on  $[0, 1]$  (Uniform).

$$\text{Need } c \geq \sup_{x \in [0,1]} \frac{\pi(x)}{q(x)} = \sup 3x^2 = 3.$$

So  $c = 3$  is optimal; your code uses  $c = 4$  (still valid, slightly less efficient).

## Exercise 4.4 (R code, part 1/4): visualise + choose $c$

```
#It helps to plot the distribution so we can get an idea of what value c to use.  
#It looks like a u[0, 1] distribution with c >=3 will work here. I'll use c=4, but 3 is the most  
efficient.  
X <- seq(0, 1, 0.01)  
plot(X, 3*X^2, type = 'l')  
  
c <- 4
```

## Exercise 4.4 (R code, part 2/4): initialise

```
x <- rep(0,10000)  
count <- 1
```

## Exercise 4.4 (R code, part 3/4): rejection loop

```
while (count < 10001){  
  #Generate from Q (uniform)  
  y <- runif(1)  
  
  # Calculate the probability density function values  
  f <- 3*y^2  
  
  #Generate acceptance probabilities  
  k <- f/(c*1)  
  
  # Accept samples based on the acceptance/rejection criterion  
  u <- runif(1)  
  if (u < k){  
    x[count] <- y  
    count <- count +1  
  }  
}
```

## Exercise 4.4 (R code, part 4/4): check vs true density

```
#Plot results  
hist(x, freq = FALSE, main = "", breaks = 20)  
lines(X, 3*X^2)
```

## Exercise 4.5: half-normal via Exp(1) proposal

**Target (half-normal):**

$$\pi(x) = \frac{2}{\sqrt{2\pi}} \exp(-x^2/2), \quad x \geq 0.$$

**Proposal:**  $q(x) = \lambda e^{-\lambda x}$ .

For  $\lambda = 1$ , the envelope constant is

$$c = \sup_{x \geq 0} \frac{\pi(x)}{q(x)} = \frac{2}{\sqrt{2\pi}} e^{1/2}.$$

(That is the value used in your code.)

## Exercise 4.5 (R code, part 1/4): plot target + envelope

```
#Let us use lambda=1  
#It helps to plot the distribution so we know that our value of c is correct  
X <- seq(0, 5, 0.01)  
plot(X, 2/(sqrt(2*pi))*exp(-X^2/2), type = 'l')  
lines(X, 2/sqrt(2*pi)*exp(1/2)*dexp(X, 1), col = 2)
```

## Exercise 4.5 (R code, part 2/4): propose + compute f

```
#Generate from Q (uniform)  
y <- rexp(10000, 1)  
  
# Calculate the probability density function values  
f <- 2/(sqrt(2*pi))*exp(-y^2/2)
```

## Exercise 4.5 (R code, part 3/4): accept/reject

```
#Generate acceptance probabilities  
c <- 2/sqrt(2*pi)*exp(1/2)  
k <- f/(c*dexp(y, 1)) #q(y) = 1  
  
# Generate random samples from a uniform distribution for acceptance/rejection  
u <- runif(10000)  
  
# Accept samples based on the acceptance/rejection criterion  
x <- y[u < k]
```

## Exercise 4.5 (R code, part 4/4): check vs true density

```
#Plot results  
hist(x, freq = FALSE, main = "", breaks = 20)  
lines(X, 2/(sqrt(2*pi))*exp(-X^2/2))
```

- Ex 4.2: normalise to get  $a = 3$ , then  $X = U^{1/3}$ .
- Ex 4.3:  $F^{-1}(u) = u^\theta$  for  $\theta \in \{1, 5, 10\}$ .
- Ex 4.4: rejection on  $[0, 1]$  with Uniform proposal;  $c = 3$  optimal (your code uses  $c = 4$ ).
- Ex 4.5: rejection with  $\text{Exp}(1)$  proposal; envelope constant  $c = \frac{2}{\sqrt{2\pi}}e^{1/2}$ .