

## Homework 3

### 3.3

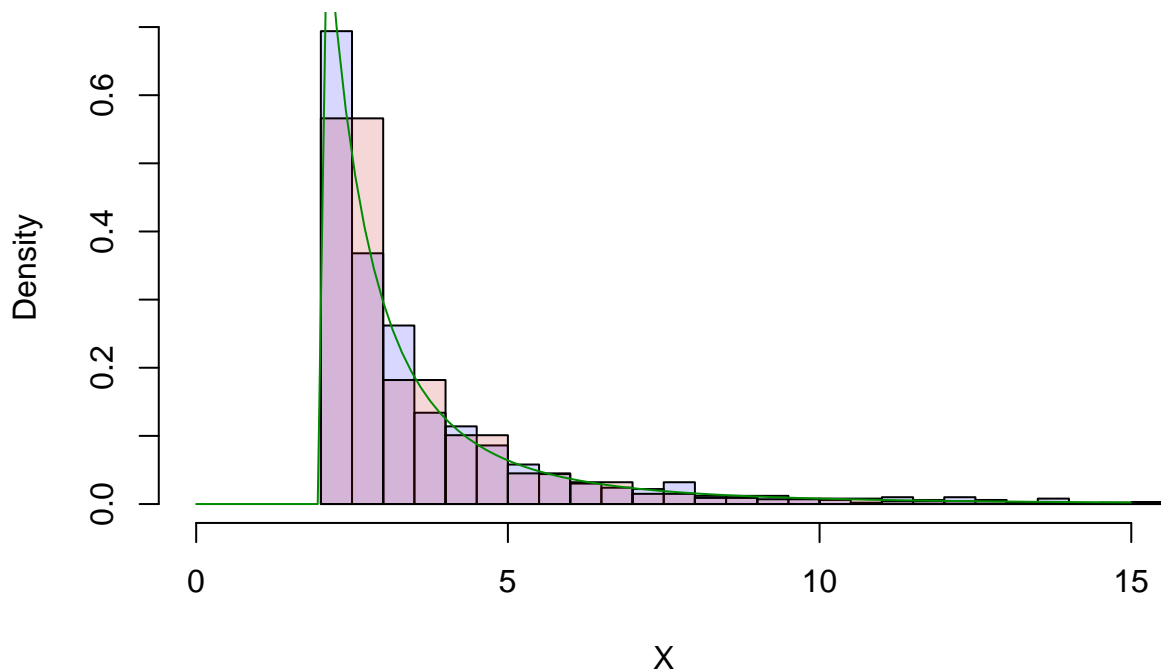
The  $Pareto(a, b)$  distribution has cdf  $F(x) = 1 - (\frac{b}{x})^a$ ,  $x \geq b > 0$ ,  $a > 0$ . Derive the probability inverse transformation  $F^{-1}(U)$  and use the inverse transformation method to simulate a random sample from the  $Pareto(2,2)$  distribution. Graph the density histogram of the sample with the  $Pareto(2,2)$  density superimposed for comparison.

Ans.

$$F^{-1}(U) = b(1 - u)^{-1/a}$$

```
set.seed(165)
n <- 1000
u <- runif(n) #Uniform Sample
a <- 2
b <- 2
X <- b*(1-u)^(-1/a) #Pareto Sample Using Inverse Transformation
p <- rpareto(n,2,2)
hist(X, freq=F, col=mycol1,breaks=200, xlim=c(0,15), main="Pareto(2,2)") #Histogram of Inverse Transform
hist(p, freq=F, col=mycol2,breaks=50, add=T, xlim=c(0,15)) #Histogram of Random Pareto Sample
curve(dpareto(x,2,2), add = T, col='green4', xlim=c(0,15)) #Density curve for Pareto(2,2)
```

### Pareto(2,2)



### 3.4

The Rayleigh density is  $f(x) = \frac{x}{\sigma^2} e^{-x^2/(2\sigma^2)}$ ,  $x \geq 0$ ,  $\sigma > 0$ . Develop an algorithm to generate random samples from a Rayleigh( $\sigma$ ) samples for several choices of  $\sigma > 0$  and check that the mode of the generated samples is close to the theoretical mode  $\sigma$  (check the histogram).

Ans.

$$F(x) = 1 - e^{-x^2/2\sigma^2} \Rightarrow F^{-1}(U) = \sigma\sqrt{-2\ln U}$$

```
raysample <- function(s, u){  
  r <- s*sqrt(-2*log(u))  
  return (r)  
}  
  
par(mfrow=c(1,3))  
  
hist(raysample(1, u), freq=F, col='blue4', main="Rayleigh(1)")  
curve(drayleigh(x,1), add=T, col='red4', lw=2)  
abline(v=1, col='green4')  
  
hist(raysample(2, u), freq=F, col='blue4', main="Rayleigh(2)")  
curve(drayleigh(x,2), add=T, col='red4', lw=2)  
abline(v=2, col='green4')  
  
hist(raysample(3, u), freq=F, col='blue4', main="Rayleigh(3)")  
curve(drayleigh(x,3), add=T, col='red4', lw=2)  
abline(v=3, col='green4')
```

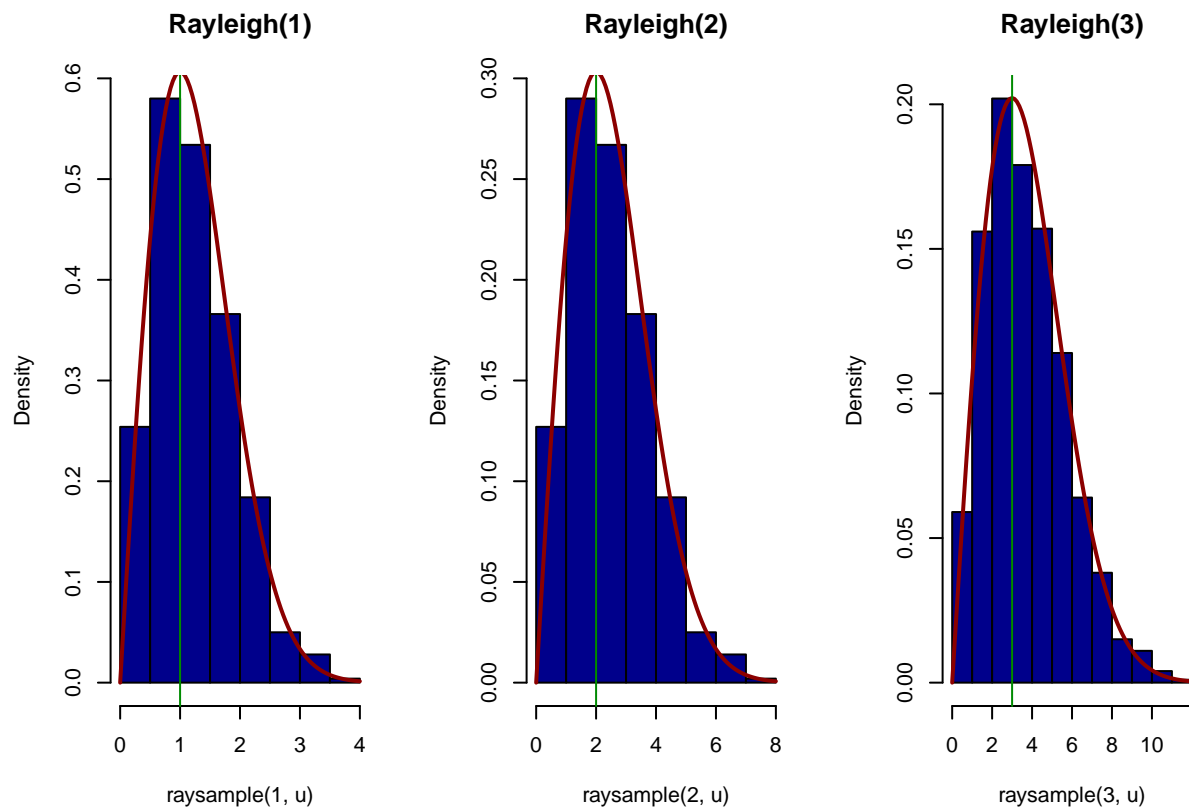


Table of the modes from the Rayleigh samples (rounded to one decimal) given different  $\sigma$  values:

$\sigma$	Mode	$\sigma$	Mode
1	0.9	5	4.7
2	1.9	6	5.7
3	2.8	7	6.6
4	3.8	8	7.6

### 3.5

A discrete random variable  $X$  has probability mass function

x	0	1	2	3	4
p(x)	0.1	0.2	0.2	0.2	0.3

Use the inverse transform method to generate a random sample of size 1,000 from the distribution of  $X$ . Construct a relative frequency table and compare the empirical with the theoretical probabilities. Repeat using the *R sample* function.

```
set.seed(165)

u2 <- runif(1000)

X2 <- numeric(1000)

for (i in 1:1000){

  if (u2[i] <= 0.1){
    X2[i] <- 0
  } else {
    if (u2[i] <= 0.3 && u2[i] >0.1){
      X2[i] <- 1
    } else{
      if( u2[i] <= 0.5 && u2[i] >0.3){
        X2[i] <- 2
      } else{
        if( u2[i] <= 0.7 && u2[i] >0.5){
          X2[i] <- 3
        } else{
          if( u2[i] <= 1 && u2[i] >0.7){
            X2[i] <- 4
          }
        }
      }
    }
  }
}

}
```

Comparative of Empirical Probabilities from the sample X and the Theoretical Probabilities:

```
df <- as.data.frame(round(table(X2)/1000,2))
df <- cbind(df, c(0.1,0.2,0.2,0.2,0.3))
colnames(df) <- c("x", "Empirical Prob.", "Theoretical Prob.")
df
```

```
##   x Empirical Prob. Theoretical Prob.
## 1 0           0.10           0.1
## 2 1           0.20           0.2
## 3 2           0.18           0.2
## 4 3           0.20           0.2
## 5 4           0.32           0.3
```

Sample from X #1 (size = 100)

```
set.seed(165)
X2_1 <- sample(X2, 100)

df1 <- as.data.frame(round(table(X2_1)/100,2))
df1 <- cbind(df1, c(0.1,0.2,0.2,0.2,0.3))
colnames(df1) <- c("x", "Empirical Prob.", "Theoretical Prob.")
df1
```

```
##   x Empirical Prob. Theoretical Prob.
## 1 0           0.08           0.1
## 2 1           0.21           0.2
## 3 2           0.21           0.2
## 4 3           0.18           0.2
## 5 4           0.32           0.3
```

Sample from X #2 (size = 250)

```
set.seed(165)
X2_2 <- sample(X2, 250)

df2 <- as.data.frame(round(table(X2_2)/250,2))
df2 <- cbind(df2, c(0.1,0.2,0.2,0.2,0.3))
colnames(df2) <- c("x", "Empirical Prob.", "Theoretical Prob.")
df2
```

```
##   x Empirical Prob. Theoretical Prob.
## 1 0           0.07           0.1
## 2 1           0.20           0.2
## 3 2           0.18           0.2
## 4 3           0.22           0.2
## 5 4           0.33           0.3
```

Sample from X #3 (size = 500)

```
set.seed(165)
X2_2 <- sample(X2, 500)
```

```
df2 <- as.data.frame(round(table(X2_2)/500,2))
df2 <- cbind(df2, c(0.1,0.2,0.2,0.2,0.3))
colnames(df2) <- c("x", "Empirical Prob.", "Theoretical Prob.")
df2
```

```
##   x Empirical Prob. Theoretical Prob.
## 1 0           0.09           0.1
## 2 1           0.21           0.2
## 3 2           0.18           0.2
## 4 3           0.20           0.2
## 5 4           0.31           0.3
```

### 3.8

Write a function to generate random variates from a Lognormal( $\mu, \sigma$ ) distribution using a transformation method, and generate a random sample of size 1,000. Compare the histogram with the Lognormal density curve given by the *dlnorm* function in *R*.

Ans. Given  $X \sim N(\mu, \sigma)$  then  $\log(X) \sim LN(\mu, \sigma)$ . Using this transformation from the *rnorm* function in *R*.

```
set.seed(165)

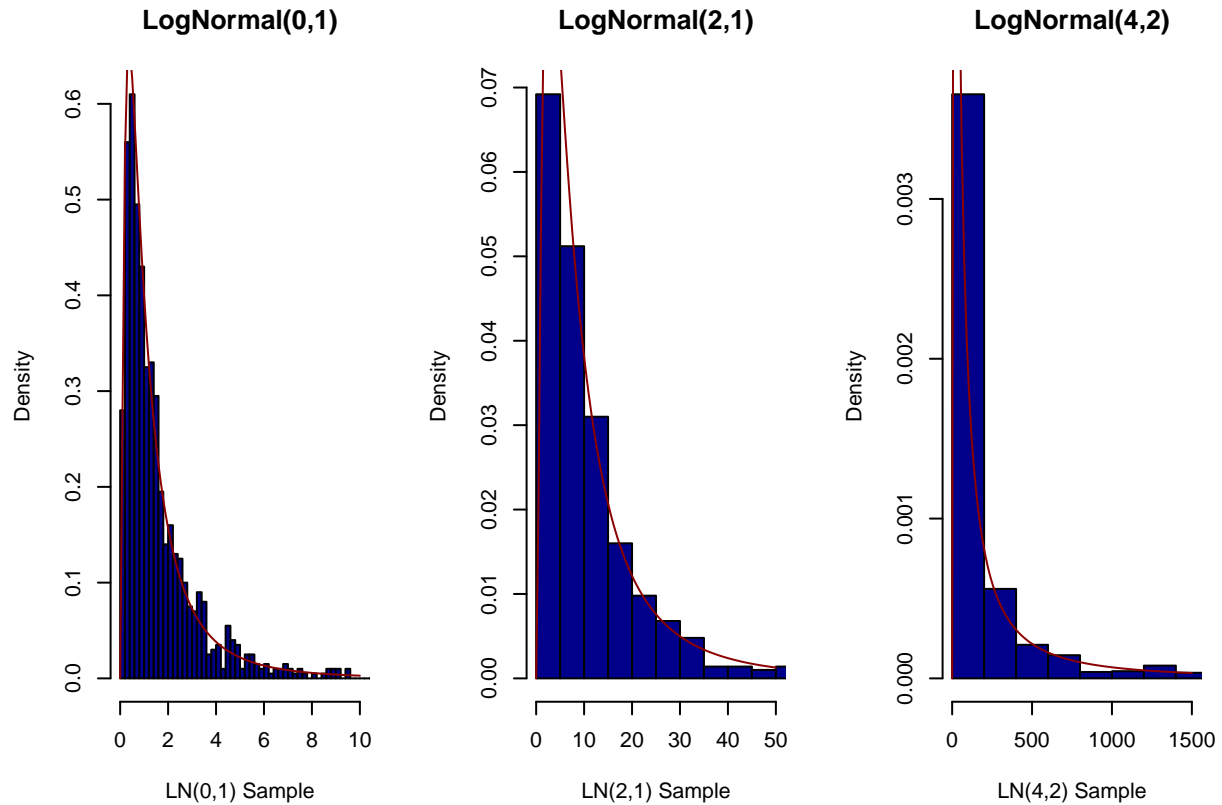
normsample <- function(mu, sigma, n){
  return (rnorm(n,mu,sigma))
}

par(mfrow=c(1,3))

hist(exp(normsample(0,1,n)), col='blue4', freq=F, breaks=100, xlim=c(0,10), xlab = "LN(0,1) Sample", ma
curve(dlnorm(x,0,1), col='red4', add=T)

hist(exp(normsample(2,1,n)), col='blue4', freq=F, breaks=100, xlim=c(0,50), xlab = "LN(2,1) Sample", ma
curve(dlnorm(x,2,1), col='red4', add=T)

hist(exp(normsample(4,2,n)), col='blue4', freq=F, breaks=250, xlim=c(0,1500), xlab = "LN(4,2) Sample", r
curve(dlnorm(x,4,2), col='red4', add=T)
```



### 3.9

The rescaled Epanechnikov kernel [92] is a symmetric density function

$$f_e(x) = \frac{3}{4}(1 - x^2), \quad |x| \leq 1$$

Devroye and Györfi give the following algorithm for simulation for this distribution. Generate *iid*  $U_1, U_2, U_3 \sim \text{Uniform}(-1, 1)$ . If  $|U_3| \geq |U_2|$  and  $|U_3| \geq |U_1|$ , deliver  $U_2$ ; otherwise deliver  $U_3$ . Write a function to generate random variates from  $f_e$ , and construct the histogram density estimate of a large simulated random sample.

```
set.seed(555)
n2 <- 10000
U1 <- runif(n2, -1, 1)
U2 <- runif(n2, -1, 1)
U3 <- runif(n2, -1, 1)
Uepa <- numeric(n2)

epa <- function(U1, U2, U3, n){
  Uepa <- numeric(n)
  for (i in 1:n) {
    if (abs(U3[i]) >= abs(U2[i]) && abs(U3[i]) >= abs(U1[i])) {
      Uepa[i] <- U2[i]
    } else {
      Uepa[i] <- U3[i]
    }
  }
}
```

```

    return (Uepa)
}

hist(epa(U1,U2,U3, n2), col='blue4', freq=F, xlim=c(-1,1), main='Epanechnikov Distribution', xlab='Epanechnikov Estimate',
curve((3/4)*(1-x^2), add=T, col='red4', lw=2)

```

