

Solutions to Selected Exercises  
for  
Braun and Murdoch's  
A First Course in Statistical Programming with R

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November 1, 2007

# Chapter 5

## Simulation

### 5.2 Generation of Pseudorandom Numbers

```
1. > x0 <- 17218
   > x <- numeric(20)
   > x[1] <- (172 * x0) %% 30307
   > for(i in 2:20){
+   x[i] <- (x[i-1] * 172) %% 30307
+ }
   > x

[1] 21717 7563 27942 17518 12703 2812 29059 27800 23401 24448 22690 23384
[13] 21524 4674 15946 15082 18009 6234 11503 8561

3. (a) > set.seed(32078)
      > runif(20,0,1)

[1] 0.25646258 0.49881767 0.52665491 0.62698163 0.80527541 0.18434520
[7] 0.51023267 0.36839051 0.17081759 0.74328879 0.01424726 0.22347861
[13] 0.36032442 0.79655794 0.01565888 0.40551753 0.65151347 0.91013292
[19] 0.77073083 0.51094885

      (b) > set.seed(32078)
           > runif(20,3,7)

[1] 4.025850 4.995271 5.106620 5.507927 6.221102 3.737381 5.040931 4.473562
[9] 3.683270 5.973155 3.056989 3.893914 4.441298 6.186232 3.062636 4.622070
[17] 5.606054 6.640532 6.082923 5.043795

      (c) > set.seed(32078)
           > runif(20,-2,2)

[1] -0.974149697 -0.004729333 0.106619657 0.507926506 1.221101642
[6] -1.262619189 0.040930690 -0.526437979 -1.316729628 0.973155177
[11] -1.943010969 -1.106085563 -0.558702309 1.186231747 -1.937364492
[16] -0.377929887 0.606053870 1.640531669 1.082923308 0.043795402

5. (a) > r <- runif(10000, 3.7, 5.8)
      > mean(r)

[1] 4.756022

      > var(r)

[1] 0.3664445

      > sd(r)

[1] 0.6053466
```

```

(b) > length(r[r>4])/length(r)
      [1] 0.8584

7. > U1 <- runif(10000)
   > U2 <- runif(10000)
   > U3 <- runif(10000)
   > U <- U1+U2+U3

(a) > mean(U)
      [1] 1.50653
(b) > var(U)
      [1] 0.2548007
   > var(U1)+var(U2)+var(U3)
      [1] 0.251535
(c) > mean(sqrt(U))
      [1] 1.207929
(d) > V <- sqrt(U1)+sqrt(U2)+sqrt(U3)
   > length(V[V>=.8])/length(V)
      [1] 0.9965

9. (a) > sample(1:100, size=50, replace=F)
      [1] 26 71 83 86 60 37 85 97 34 33 25 91 78 15 74 69 84 82 63 64 35 40 73 38 16
      [26] 32 22 43 19 88 68 12 8 80 17 81 99 4 66 79 49 11 31 1 61 7 46 27 39 9

(b) > sample(1:100, size=50, replace=T)
      [1] 80 32 67 86 77 87 48 41 85 53 29 17 76 72 66 56 32 36 17 61 27 87 8 2 18
      [26] 42 88 99 60 80 38 25 68 79 98 9 40 34 16 72 18 87 15 65 8 65 3 90 15 64

```

## 5.3 Simulation of Other Random Variables

### 5.3.1 Bernoulli random variables

- (a) 

```
> r <- rbinom(n=10, size=1, p=.5)
> sum(r)
[1] 6
```

(b) 

```
> r <- rbinom(n=1000, size=1, p=.5)
> sum(r)
[1] 481
```
- ```
> r <- rbinom(n=500, size=1, p=.99)
> mean(r)
[1] 0.99
> var(r)
[1] 0.00991984
```

Theoretical values are .99 and .0099.

### 5.3.2 Binomial random variables

- ```
> rbinom(p=0.15, size=25, n=24)
[1] 5 4 5 5 4 2 3 3 1 2 4 2 3 2 5 5 4 5 3 3 3 8 5 4
> # Exceeds 5 once.
> rbinom(p=0.2, size=25, n=24)
[1] 5 6 5 5 4 8 8 4 5 7 7 5 7 3 6 4 5 4 4 6 6 6 7 9
> # Exceeds 5 several times.
> rbinom(p=0.25, size=25, n=24)
[1] 6 3 1 13 5 11 11 3 4 7 9 4 4 1 5 6 6 7 9 6 9 7 7 9
```
- ```
> r <- rbinom(n=1000, size=18, p=.76)
> mean(r)
[1] 13.697
> var(r)
[1] 3.472664
```

Theoretical values are 13.68 and 3.2832.

- (a) 

```
> #Generate binomial pseudorandom variables using inversion method
> ranbin2 <- function(n, size, prob){
+   #'singlenuumber' generates one binomial random variable.
+   singlenuumber <- function(size, prob){
+     x <- runif(size)
+     N <- sum(x<prob)
+     N
+   }
+   replicate(n, singlenuumber(size, prob))
+ }
```

(b) 

```
> system.time(gcFirst=T, ranbin(n=10000, size=10, prob=.4))
```

```

user system elapsed
0.15 0.00 0.16
> system.time(gcFirst=T, rbinom(n=10000, size=10, prob=.4))
user system elapsed
0.02 0.00 0.02
> system.time(gcFirst=T, ranbin(n=10000, size=100, prob=.4))
user system elapsed
0.18 0.00 0.19
> system.time(gcFirst=T, rbinom(n=10000, size=100, prob=.4))
user system elapsed
0 0 0
> system.time(gcFirst=T, ranbin(n=10000, size=1000, prob=.4))
user system elapsed
0.33 0.00 0.33
> system.time(gcFirst=T, rbinom(n=10000, size=1000, prob=.4))
user system elapsed
0.01 0.00 0.02

```

### 5.3.3 Poisson random variables

```

1. > rpois(n=15, lambda=2.8)
[1] 1 2 4 3 3 4 2 5 4 2 4 1 2 2 1

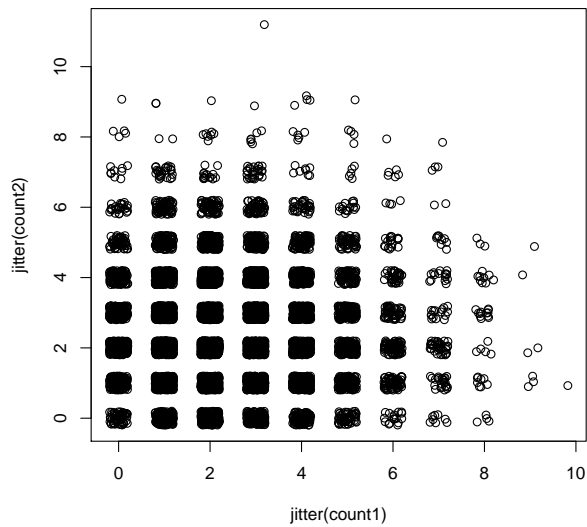
3. > x <- rpois(10000, lambda=7.2)
> mean(x)
[1] 7.141
> var(x)
[1] 7.287248

7. > N <- rpois(10000, lambda=2.5 * 2)

(a) > count1 <- c()
> count2 <- c()
> for(i in 1:10000){
+   u <- runif(N[i], max=2)
+   count1[i] <- sum(u<1)
+   count2[i] <- sum(u>1)
+ }
> sum(count1)
[1] 24886
> sum(count2)
[1] 24889

(b) yes
(c) > plot(jitter(count1), jitter(count2))

```



### 5.3.4 Exponential random numbers

```

1. > r <- rexp(50000, rate=3)

  (a) > sum(r<1)/length(r)
      [1] 0.95204
      > pexp(1, rate=3)
      [1] 0.950213

  (b) > mean(r)
      [1] 0.3309469

  (c) > var(r)
      [1] 0.1083953

3. > r1 <- rexp(100000, rate=1/3)
   > r2 <- rexp(100000, rate=1/6)
   > index <- (r2-r1)>0
   > r.min <- c(r1[index], r2[!index])
   > mean(r.min)
      [1] 1.984397
   > var(r.min)
      [1] 3.961419

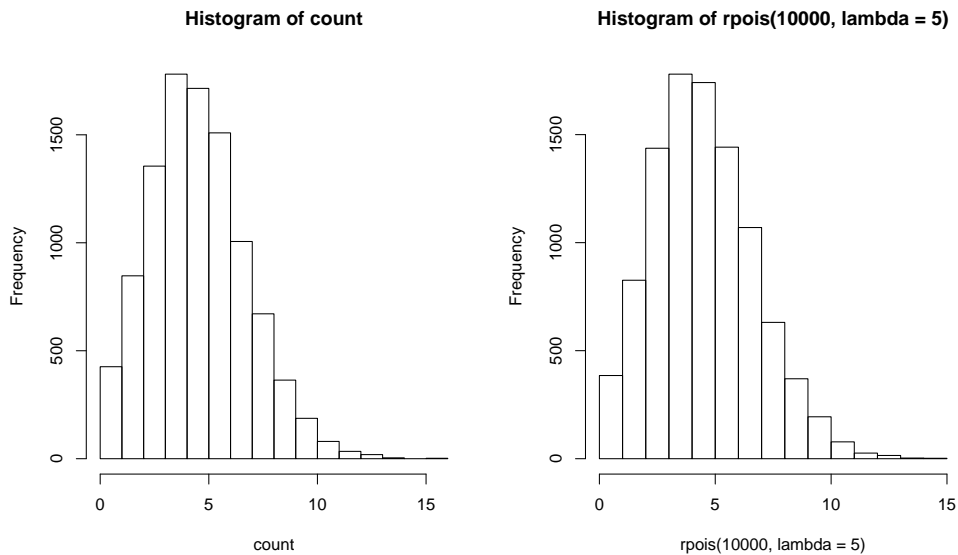
5. > count <- c()
   > for(i in 1:10000){
+   cum <- 0
+   j <- 0
+   while(cum<=2){
+     j <- j+1
+     r <- rexp(1, rate=2.5)
+     cum <- cum + r
+   }

```

```

+   count[i] <- j - 1
+ }
> par(mfrow=c(1,2))
> hist(count, breaks=20)
> hist(rpois(10000, lambda=5), breaks=20)

```



A QQ-plot would also be an appropriate way of checking this.

### 5.3.5 Normal random variables

```

1. > r <- rnorm(n=100, mean=51, sd=5.2)
   > mean(r)
[1] 51.63488
   > sd(r)
[1] 4.797961
3. > sim <- function(){
+   ans <- 0
+   while(abs(ans)<=2){
+     ans <- rnorm(1, mean=3, sd=4)
+   }
+   ans
+ }
> sim()
[1] 7.188694
5. > r <- rchisq (n=100, df=8)
   > mean(r)
[1] 8.36974
   > var(r)
[1] 18.08655

```

## 5.4 Monte Carlo Integration

```
1. > #Monte Carlo
> u <- runif(1000)
> mean(u)

[1] 0.5140157

> #Integrate function
> f <- function(x){
+   x
+ }
> integrate(f, lower=0, upper=1)

0.5 with absolute error < 5.6e-15

> #Monte Carlo
> u <- runif(1000, max=3)
> mean(u^2)*(3-0)

[1] 9.138817

> #Integrate function
> f <- function(x){
+   x^2
+ }
> integrate(f, lower=0, upper=3)

9 with absolute error < 1e-13

> #Monte Carlo
> u <- runif(1000, max=pi)
> mean(sin(u))*(pi-0)

[1] 1.988680

> #Integrate function
> f <- function(x){
+   sin(x)
+ }
> integrate(f, lower=0, upper=pi)

2 with absolute error < 2.2e-14

> #Monte Carlo
> u <- runif(1000, min=1, max=pi)
> mean(exp(u)/dunif(u,min=1, max=pi))

[1] 20.51446

> #Integrate function
> f <- function(x){
+   exp(x)
+ }
> integrate(f, lower=1, upper=pi)

20.42241 with absolute error < 2.3e-13

> #Monte Carlo
> r <- rexp(1000)
> mean(exp(-r)/dexp(r))

[1] 1
```



```

> #Integrate function
> f <- function(x){
+   exp(-x)
+ }
> integrate(f, lower=0, upper=1000)

1 with absolute error < 9e-10

> #Monte Carlo
> r <- rexp(1000)
> mean(exp(-r^3)/dexp(r))

[1] 0.9187494

> #Integrate function
> f <- function(x){
+   exp(-x^3)
+ }
> integrate(f, lower=0, upper=1000)

0.8929795 with absolute error < 8.4e-10

> #Monte Carlo
> r <- runif(1000, max=3)
> mean(sin(exp(r))/dunif(r, max=3))

[1] 0.617729

> #Integrate function
> f <- function(x){
+   sin(exp(x))
+ }
> integrate(f, lower=0, upper=3)

0.6061245 with absolute error < 2.8e-10

> #Monte Carlo
> r <- runif(1000,max=1)
> mean(dnorm(r))

[1] 0.3382709

> #Integrate function
> f <- function(x){
+   dnorm(x)
+ }
> integrate(f, lower=0, upper=1)

0.3413447 with absolute error < 3.8e-15

> #Monte Carlo
> r <- runif(1000,max=2)
> mean(dnorm(r)*2)

[1] 0.4830885

> #Integrate function
> f <- function(x){
+   dnorm(x)
+ }
> integrate(f, lower=0, upper=2)

0.4772499 with absolute error < 5.3e-15

```

```

> #Integrate function
> f <- function(x){
+   dnorm(x)
+ }
> integrate(f, lower=0, upper=3)

0.4986501 with absolute error < 5.5e-15

```

## 5.5 Advanced Simulation Methods

### 5.5.2 Importance Sampling

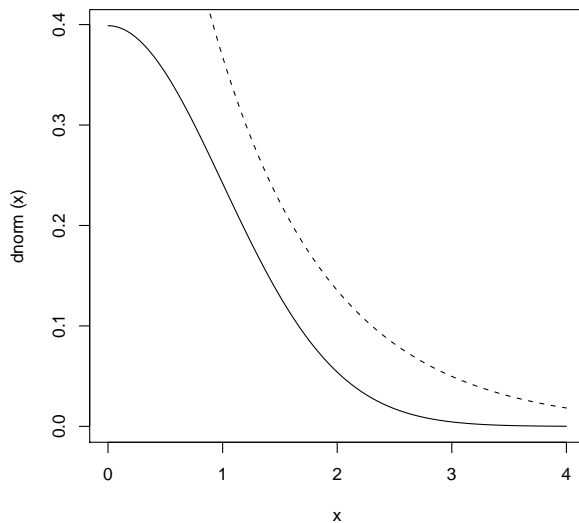
```

1. > rand.normal <- function(n){
+   r1 <- runif(n, min=-4, max=4)
+   r2 <- runif(n)
+   ans <- r1[ r2 < 2.5*dnorm(r1)]
+   while(length(ans) < n){
+     r1 <- runif(n, min=-4, max=4)
+     r2 <- runif(n)
+     ans <- c(ans,r1[ r2 < 2.5*dnorm(r1)])
+   }
+   ans[1:n]
+ }

3. > r <- rannorm(10000,8,2)

5. > curve(dnorm, from =0, to =4)
> curve(dexp, from =0, to =4, add=T, lty=2)

```



```

7. > set.seed(91626)
> probs <- runif(100)
> probs <- probs/sum(probs)
> system.time(gcFirst=T, r <- randiscrete1(100, probs))

user system elapsed
0      0      0

```

```

> system.time(gcFirst=T, r <- randiscrete1(1000, probs))
  user  system elapsed
 0.02   0.00   0.02
> system.time(gcFirst=T, r <- randiscrete1(10000, probs))
  user  system elapsed
  0.2   0.0   0.2
> system.time(gcFirst=T, r <- randiscrete2(100, probs))
  user  system elapsed
  0     0     0
> system.time(gcFirst=T, r <- randiscrete2(1000, probs))
  user  system elapsed
 0.06   0.00   0.06
> system.time(gcFirst=T, r <- randiscrete2(10000, probs))
  user  system elapsed
 0.69   0.00   0.69

```

When inversion method is not available.

## 5.6 Chapter Exercises

```

1. > simulate <- function(AnnFirst = T, pAnn, pBob){
+   scoreAnn <- 0
+   scoreBob <- 0
+   nextPlayer <- 'Ann'
+   if(AnnFirst==F){
+     nextPlayer <- Bob
+   }
+   while(scoreAnn < 21 & scoreBob < 21){
+     if(nextPlayer=='Ann'){
+       r <- rbinom(1, 1, pAnn)
+       if(r==1){
+         scoreAnn <- scoreAnn + 1
+       }else{
+         nextPlayer <- 'Bob'
+       }
+     }else{
+       r <- rbinom(1, 1, pBob)
+       if(r==1){
+         scoreBob <- scoreBob + 1
+       }else{
+         nextPlayer <- 'Ann'
+       }
+     }
+   }
+   return(nextPlayer)
+ }
> pAnn <- .7
> pBob <- .72
> result <- c()
> for(i in 1:7){

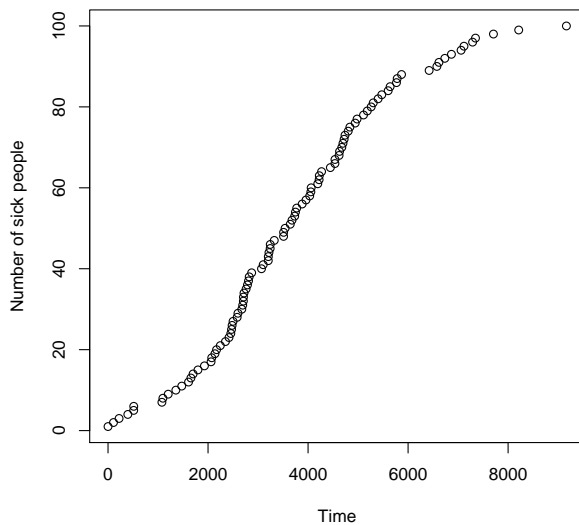
```

```

+   result <- c(result,simulate(AnnFirst=T, pAnn=pAnn, pBob=pBob))
+ }

2. > simulate <- function(N, p){
+   time.line <- 0
+   time.current <- 0
+   n.sick <- 1
+
+   while(n.sick<N){
+     time.current <- time.current + 1
+
+     #if encounters==0, then none of the two persons is sick
+     #if encounters==1, then one of the two persons is sick
+     #if encounters==2, then both of the two persons are sick
+     encounters <- rhyper(nn=1, m=n.sick, n=N-n.sick, k=2)
+
+     if(encounters==1){
+       contage <- rbinom(n=1, size=1, p)
+       if(contage==1){
+         #record current time
+         time.line <- c(time.line, time.current)
+         n.sick <- n.sick + 1
+       }
+     }
+   }
+   time.line
+ }
> N <- 100
> p <- .05
> ans <- simulate(N, p)
> plot(ans, 1:N, xlab='Time', ylab='Number of sick people')

```



```

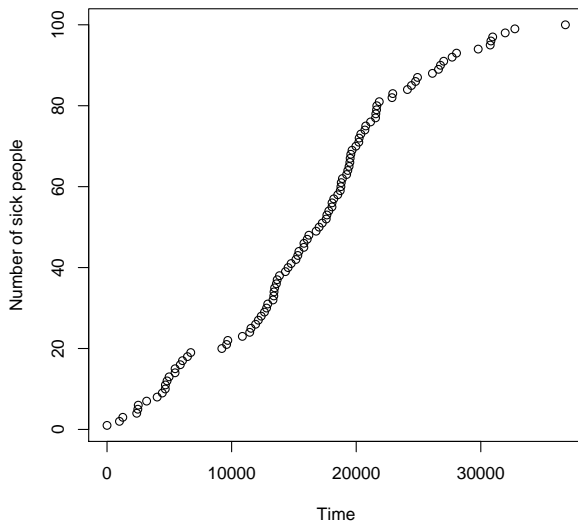
5. > simulate <- function(N, p){
+   time.line <- 0
+   time.current <- 0

```

```

+   n.sick <- 1
+
+   while(n.sick<N){
+     time.current <- time.current + rexp(n=1,rate=1/5)
+
+     #if encounters==0, then none of the two persons is sick
+     #if encounters==1, then one of the two persons is sick
+     #if encounters==2, then both of the two persons are sick
+     encounters <- rhyper(nn=1, m=n.sick, n=N-n.sick, k=2)
+
+     if(encounters==1){
+       contage <- rbinom(n=1, size=1, p)
+       if(contage==1){
+         #record current time
+         time.line <- c(time.line, time.current)
+         n.sick <- n.sick + 1
+       }
+     }
+   }
+   time.line
+ }
> N <- 100
> p <- .05
> ans <- simulate(N, p)
> plot(ans, 1:N, xlab='Time', ylab='Number of sick people')

```



```

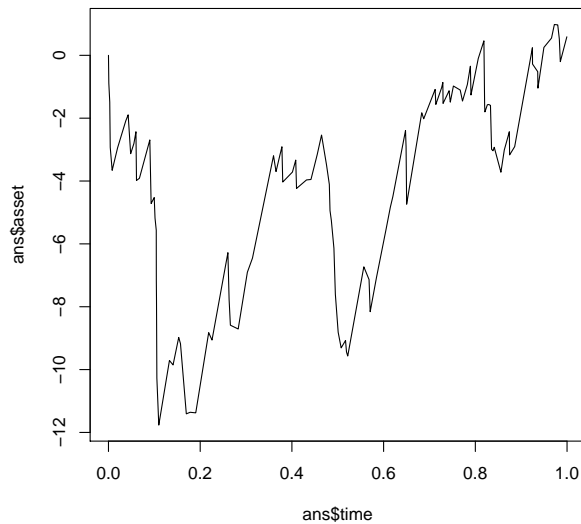
7. (a) > simulate <- function(){
+   N <- rpois(n=1, lambda=100)
+   claimSize <- rgamma(N, shape=2, rate=2)
+   claimTime <- sort(runif(N))
+
+   asset <- 0
+   asset.1 <- 105 * claimTime[1] - claimSize[1]
+   asset <- c(asset, asset.1)

```

```

+
+   for(i in 2:N){
+     len <- length(asset)
+     asset.i <- asset[len] + (claimTime[i] - claimTime[i-1])*105 - claimSize[i]
+     asset <- c(asset, asset.i)
+   }
+   claimTime <- c(0, claimTime)
+   list(time=claimTime, asset=asset)
+ }
> ans <- simulate()
> plot(ans$time, ans$asset, type='l')

```



```

(b) > minimum <- c()
> final <- c()
> for(i in 1:1000){
+   ans <- simulate()
+   minimum <- c(minimum, min(ans$asset))
+   final <- c(final, rev(ans$asset)[1])
+ }
> mean(minimum)
[1] -6.723402
> mean(final)
[1] 4.721359

```