

Documentation: Assignment 5

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Q 1 Generate 1000 standard normal variates using standard Double-exponential distribution by acceptance-rejection method. Calculate the necessary constant c , where

$$f(x)/g(x) \leq c$$

$$f(x) \text{ and } g(x)$$

are the pdfs of standard normal and standard Double-exponential distribution respectively. Calculate the theoretical and simulated acceptance probability. How do you justify your generated random numbers are correct? Provide as many verification as you can.

Code: R

```
1 count<-0
2 gInv<-function(u){
3   if(u<1/2){
4     return(log(2*u))
5   }
6   else{
7     return(-1*log(2*(1-u)))
8   }
9 }
10 c=sqrt(2*exp(1)/pi)
11 f<-function(x){
12   return(sqrt(1/(2*pi))*exp(-1*x*x/2))
13 }
14 g<-function(x){
15   return((1/2)*exp(-1*abs(x)))
16 }
17 vector<-c()
18 total<-0
19 while(1){
20   u1<-runif(1)
21   u2<-runif(1)
22   x<-gInv(u1)
23   if(u2<=(f(x)/(c*g(x)))){
24     cat("Sample Values are: ",x)
25     cat("\n")
26     count<-count+1
27     vector[count]<-x
28   }
```

```
29 total<-total+1
30 if (count==1000)
31 break
32 }
33 cat("\nSimulated Acceptance: ",count/total)
34 cat("\nTheoretical Acceptance: ",1/c)
35 cat("\nMean: ",mean(vector))
36 cat("\nVariance: ",var(vector))
37 png("question1.png")
38 hist(vector, breaks=50, col="light cyan", plot=TRUE)
39 cat("\n")
```

Output:

Simulated Acceptance: 0.7745933

Theoretical Acceptance: 0.7601735

Mean: 0.04431985

Variance: 0.9399264

Observation:

$$g(x) = 1/2 * e^{(-|x|)} \quad (1)$$

$$f(x) = \sqrt{1/2\pi} * e^{(-x^2/2)} \quad (2)$$

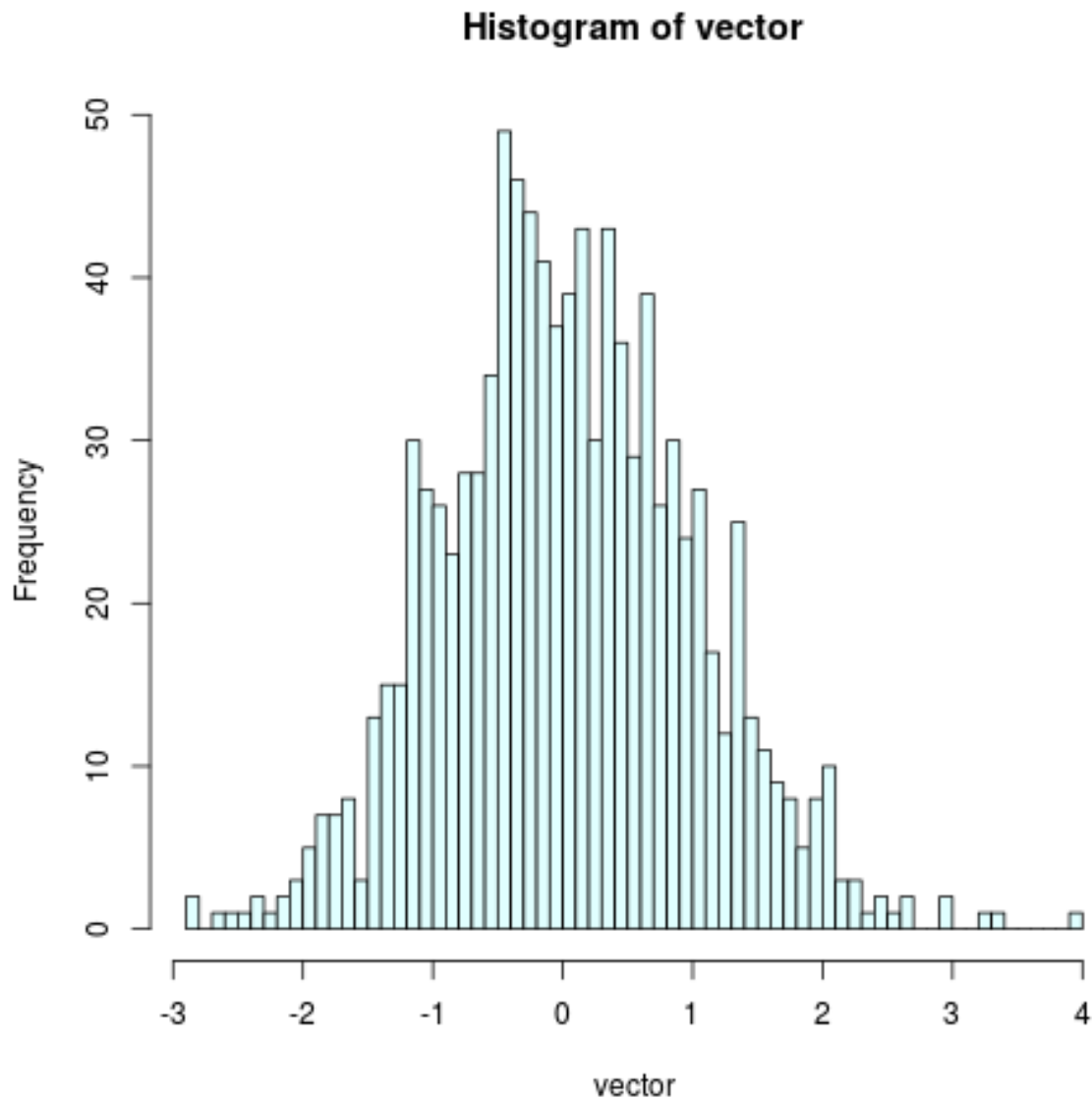
$$f(x)/g(x) = \sqrt{2/\pi} e^{(-x^2/2+|x|)} \leq \sqrt{2e/\pi} = \text{constant} \quad (3)$$

- Simulated acceptance probability: 0.7745933
- Theoretical acceptance probability: 0.7601735
- **Justification of why my random numbers are correct:** As my mean and variance are closer to the standard normal function's respective attributes and also graph is similar to it. So my output is near to correct distribution.

Mean: 0.04431985

Variance: 0.9399264

Graph:



(a) using R

Q 2 Do the same exercise for generating random numbers from half-standard normal distribution using exponential distribution with mean 1 by acceptance-rejection method.

Code: R

```
1 count<-0
2 gInv<-function(u){
3   return(-1*log(1-u))
4 }
5 c=sqrt(2*exp(1)/pi)
6 f<-function(x){
```

```
7   return (2*sqrt(1/(2*pi))*exp(-1*x*x/2))
8 }
9 g<-function(x){
10   return (exp(-1*x))
11 }
12 vector<-c()
13 total<-0
14 while(1){
15   u1<-runif(1)
16   u2<-runif(1)
17   x<-gInv(u1)
18   if(u2<=(f(x)/(c*g(x)))) {
19     cat("Sample Values are: ",x)
20     cat("\n")
21     count<-count+1
22     vector[count]<-x
23   }
24   total<-total+1
25   if(count==10000)
26     break
27 }
28 cat("\nSimulated Acceptance: ",count/total)
29 cat("\nTheoretical Acceptance: ",1/c)
30 cat("\nMean: ",mean(vector))
31 cat("\nVariance: ",var(vector))
32 png("question2.png")
33 hist(vector, breaks=50, col="light cyan", plot=TRUE)
34 cat("\n")
```

Output:

Simulated Acceptance: 0.7628347

Theoretical Acceptance: 0.7601735

Mean: 0.8043221

Variance: 0.367382

Observation:

$$g(x) = e^{(-x)} \quad (4)$$

$$f(x) = 2\sqrt{1/2\pi} * e^{(-x^2/2)} \quad (5)$$

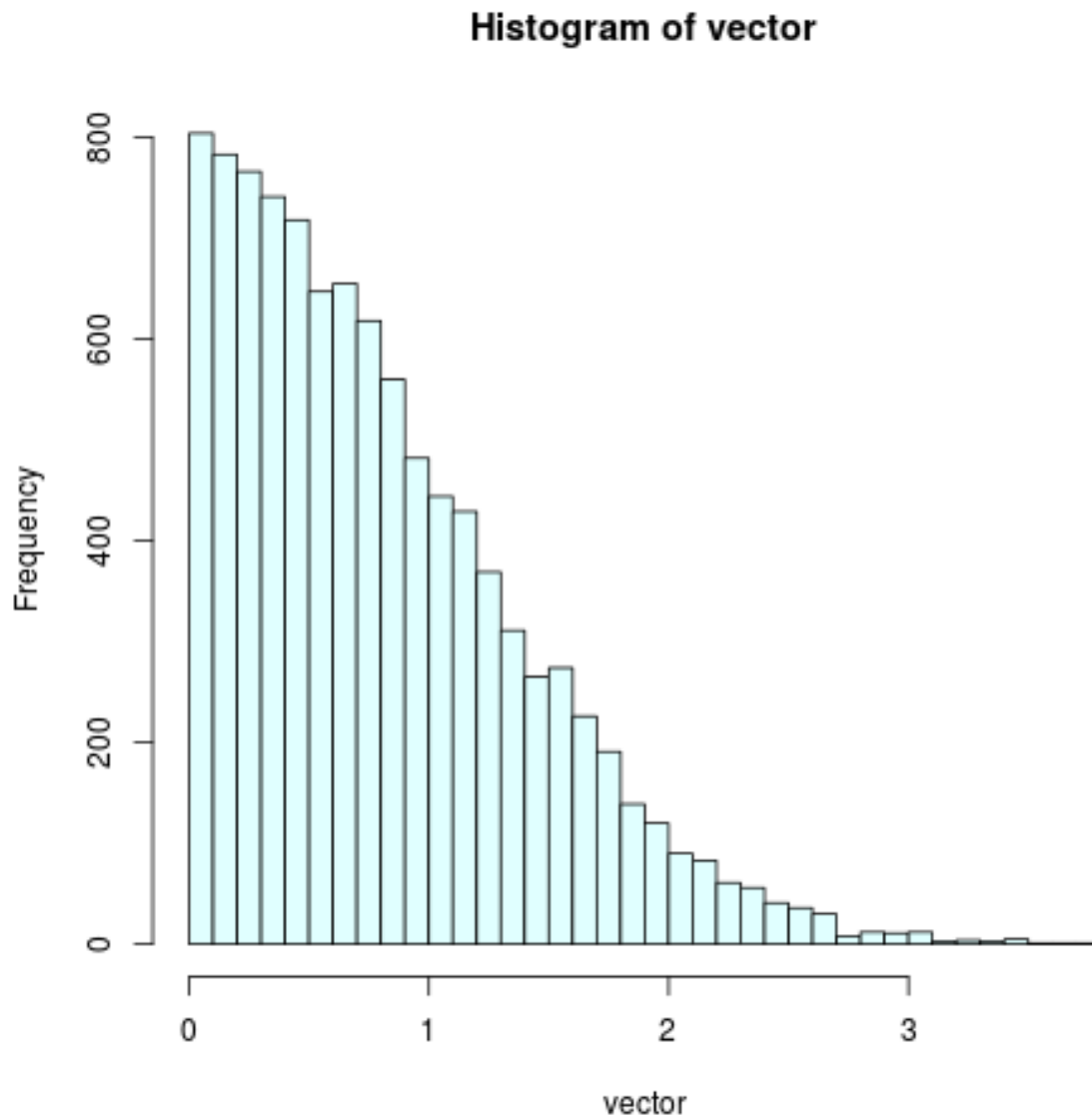
$$f(x)/g(x) \leq \sqrt{2e/\pi} = \text{constant} \quad (6)$$

- Simulated acceptance probability: 0.7628347
- Theoretical acceptance probability: 0.7601735
- **Justification of why my random numbers are correct:** As my mean and variance are closer to the standard half normal function's respective attributes and also graph is similar to it. So my output is near to correct distribution.

Mean: 0.8043221

Variance: 0.367382

Graph:



(b) Using R

Q 3 Consider the following discrete distribution.

- a) Generate 10 random numbers from the above probability mass function using usual procedure (inverse transform) of generating random number from discrete distribution defined on finite number of points. Calculate mean and variance of the generated numbers.
- b) Generate 10 random numbers from the same probability mass function by acceptance-rejection principle. Calculate mean and variance of the generated numbers

Code for part "A": Inverse Transform

```
1 cdf<-c(0.05,0.30,0.75,0.90,1.0)
2 values<-c(1,2,3,4,5)
3 samples<-c()
4 count<-0
5 while(count!=10){
6   u<-runif(1)
7   for(i in 1:5){
8     if(u<=cdf[i]){
9       count<-count+1
10      samples[count]=values[i]
11      cat("\nSample Values are: ",samples[count])
12      break
13    }
14  }
15 }
16 cat("\nMean: ",mean(samples))
17 cat("\nVariance: ",var(samples))
18 cat("\n")
```

Output of Inverse Transform generated from 10 values:

Mean: 3.1

Variance: 1.433333

Code for part "B": Acceptance Rejection

```
1 pdf<-c(0.05,0.25,0.45,0.15,0.10)
2 values<-c(1,2,3,4,5)
3 samples<-c()
4 N<-5
5 count<-0
6 max<-0
7 gInv<-function(u){
8   return(floor(N*u)+1)
9 }
10 for(i in 1:N){
11   if(max<=(pdf[i]*N)) {
12     max<-pdf[i]*N
13   }
14 }
15 f<-function(x){
```



```
16     return (pdf[x])
17 }
18 g<-function() {
19     return (1/N)
20 }
21 while(1) {
22     u1<-runif(1)
23     u2<-runif(1)
24     x<-gInv(u1)
25     if(u2<=(f(x)/(max*g()))){
26         cat("Sample Values are: ",x)
27         cat("\n")
28         count<-count+1
29         samples[count]<-x
30     }
31     if(count==10)
32         break
33 }
34 cat("\nMean: ",mean(samples))
35 cat("\nVariance: ",var(samples))
36 cat("\n")
```

Output of Acceptance Rejection generated from 10 values:

Mean: 3.2

Variance: 0.8444444