Assignment-4

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1 Question 1

Code for R

```
1 n <- 1000
                 #No of values generated
 2 c \leftarrow ((2 * exp(1)) / pi)^{(1/2)}
 3 \times - \mathbf{runif}(n)
 4 g <- vector(,n)
 5 \mid u \leftarrow runif(n)
 7 for (i in 1:n) {
 8 | if (x[i] < 1/2)  {
 g[i] <- x[i]
10 | x[i] \leftarrow log(2*x[i])
11 } else {
12 g[i] \leftarrow 1 - x[i]
13 x[i] < -log(2*(1-x[i]))
15 }
16
17 f \leftarrow ((1 / (2 * pi)) ^ (1/2)) * (exp(-(x^2)/2))
  rd \leftarrow x[(g*u*c) < f]
19
20
21 cat("Acceptance probability, Theoretical = ", 1/c, ", Stimulated = ", length(rd)/n, "\n")
22 cat ("Mean, Theoretical = 0", ", Stimulated = ", mean(rd), "\n")
23 cat("Median, Theoretical = 0", "Stimulated = ", median(rd), "\n")
24 \#cat("Mode, Theoretical = 0", "Stimulated = ", mode(rd), "\n")
25 cat("Variance, Theoretical = 1", ", Stimulated = ", var(rd), "\n")
26 cat ("Standard Deviation, Theoretical = 1", ", Stimulated = ", sd(rd), "\n")
27
28 hist (rd, main="Standard Normal Distribution", xlab="Range of random numbers", ylab="Density")
29 dev.copy(png,"plot1.png");
30 dev. off ();
```

```
31 | \mathbf{rm}(\mathbf{list} = \mathbf{ls}())
```

We see that the random numbers generated have nearly same/approaching probability distribution, acceptance probability, mean, median, variance and standard deviation, to Standard Normal Distribution. So, we justify that generated random numbers are correct. For example, a sample of values generated have the values:

```
Acceptance probability, Theoretical = 0.7601735, Stimulated = 0.763 Mean, Theoretical = 0, Stimulated = -0.00010277
```

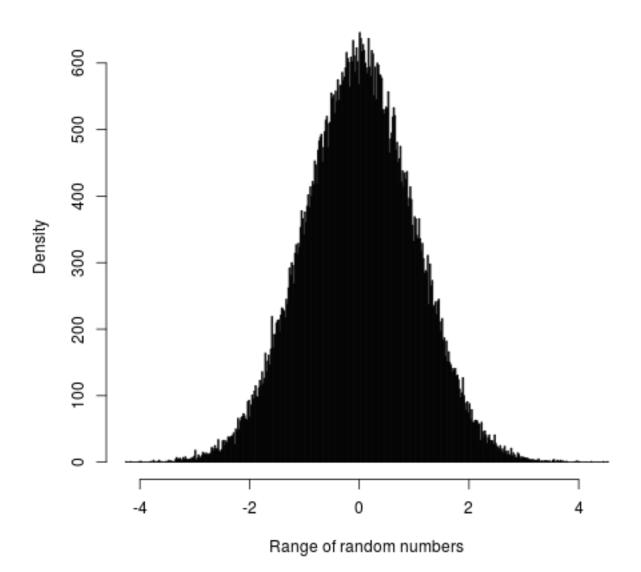
Median, Theoretical = 0 Stimulated = -0.00038851

Variance, Theoretical = 1, Stimulated = 0.9805156

Standard Deviation, Theoretical = 1, Stimulated = 0.9902099

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Standard Normal Distribution



2 Question 2

Code for R

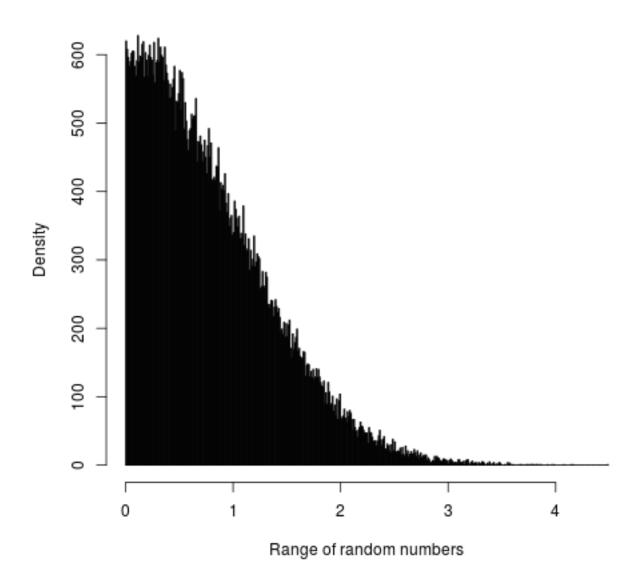
```
1 n <- 1000
                  #No of values generated
 2 | c \leftarrow ((2 * exp(1)) / pi)^{(1/2)}
 3 \times - runif(n)
4 \times -\log(x)
 5 \mid u \leftarrow runif(n)
  g \leftarrow exp(-x)
  f \leftarrow ((2/pi)^{(1/2)} + (exp(-(x^2)/2))
10 rd <- x[(g*c*u) < f]
11
12 cat("Acceptance probability, Theoretical = ", 1/c, ", Stimulated = ", length(rd)/n, "\n")
13 cat ("Mean, Theoretical = ", (2/pi)^(1/2), ", Stimulated = ", mean(rd), "\n")
14 cat ("Variance, Theoretical = ", 1 - (2/pi), ", Stimulated = ", var(rd), "\n")
15 cat ("Standard Deviation, Theoretical = ", (1 - 2/pi)^(1/2), ", Stimulated = ", sd(rd), "\n")
17 hist (rd, main="Standard Half Normal Distribution", xlab="Range of random numbers", ylab="
        Density")
18 dev.copy(png,"plot2.png");
19 dev. off ();
20 | \mathbf{rm}(\mathbf{list} = \mathbf{ls}())
```

We see that the random numbers generated have nearly same/approaching probability distribution, acceptance probability, mean, median, variance and standard deviation, to Standard Half-Normal Distribution. So, we justify that generated random numbers are correct. For example, a sample of values generated have the values:

```
Acceptance probability, Theoretical = 0.7601735, Stimulated = 0.762 Mean, Theoretical = 0.7978846, Stimulated = 0.7892713 Variance, Theoretical = 0.3633802, Stimulated = 0.3666513 Standard Deviation, Theoretical = 0.6028103, Stimulated = 0.6055174
```

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Standard Half Normal Distribution



3 Question 3

Code for R, part a

```
1 #Discrete Inverse transformation method
              #No of values generated, taking 1000 as 10 is too small for any measurement
 3 p \leftarrow c(0.05, 0.25, 0.45, 0.15, 0.10)
 4 \times (-c(1:5))
   u \leftarrow runif(n)
 7 rd <- vector(,n)
 9 for (i in 1:n) {
10 for (j in 1:5) {
11 if (((sum(p[1:j]) - p[j]) < u[i]) & (u[i] < sum(p[1:j]))) 
12 | rd[i] = j
13 }
14 }
15 }
17 print (rd)
18
19 |cat("Mean, Theoretical = ", sum(p*x), ", Stimulated = ", mean(rd), "\n")
20 |\operatorname{cat}("\operatorname{Variance}, \operatorname{Theoretical} = ", (\operatorname{sum}(p*x^2) - 9), ", \operatorname{Stimulated} = ", \operatorname{var}(\operatorname{rd}), "\setminus n")
21 hist (rd, main="Discrete Distribution", xlab="Range of random numbers", ylab="Density")
22 dev.copy(png,"plot3a.png");
23 dev. off ();
24 | \mathbf{rm}(\mathbf{list} = \mathbf{ls}())
```

Sample of 10 generated numbers are:

```
1, 3, 2, 4, 2, 3, 5, 2, 3, 3
```

Mean, Theoretical = 3, Stimulated = 2.8

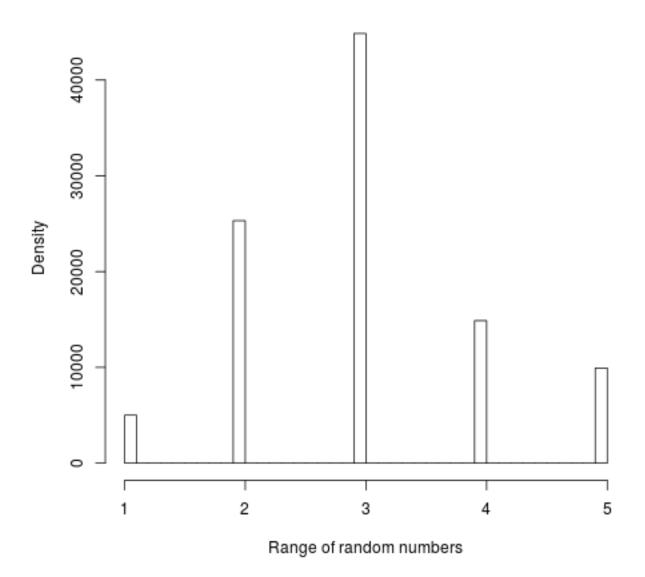
Variance, Theoretical = 1, Stimulated = 1.288889

For a sample of 1000 numbers generated, the values are very close to theoretical.

Mean, Theoretical = 3, Stimulated = 2.963

Variance, Theoretical = 1, Stimulated = 0.9946256

Discrete Distribution (for 100000 values)



Code for R, part b

```
#Discrete acceptance rejection method

n <- 20  #No of values generated, taking 1000 as 10 is too small for any measurement

p <- c(0.05, 0.25, 0.45, 0.15, 0.10)

x <- c(1:5)

c <- max(p)/0.2

u <- runif(n)

v <- runif(n)

y <- as.integer(5*u) + 1

rd <- y[ v*0.45 < p[y] ]

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```

```
print(rd)
print(rd)

cat("Mean, Theoretical = ", sum(p*x), ", Stimulated = ", mean(rd), "\n")

cat("Variance, Theoretical = ", (sum(p*x^2) - 9), ", Stimulated = ", var(rd), "\n")

hist(rd, main="Discrete Distribution", xlab="Range of random numbers", ylab="Density")

dev.copy(png, "plot3b.png");

dev.off ();

rm(list = ls())
```

Sample of 10 generated numbers are:

```
2, 5, 4, 3, 3, 2, 5, 3, 3, 2
```

Mean, Theoretical = 3, Stimulated = 3.2

Variance, Theoretical = 1, Stimulated = 1.288889

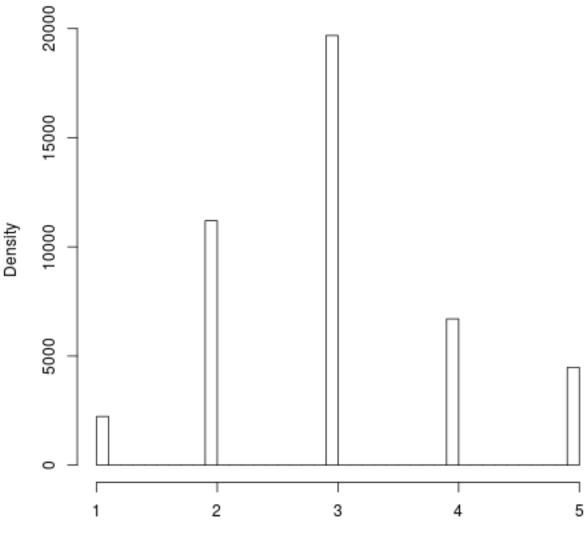
For a sample of 1000 numbers generated, the values are very close to theoretical.

Mean, Theoretical = 3, Stimulated = 3.055556

Variance, Theoretical = 1, Stimulated = 1.008675

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Discrete Distribution (for about 100000 values)



Range of random numbers