

# STAT 812: Computational Statistics

## Random Number Generator and Monte Carlo

Longhai Li

2024-09-16

## Contents

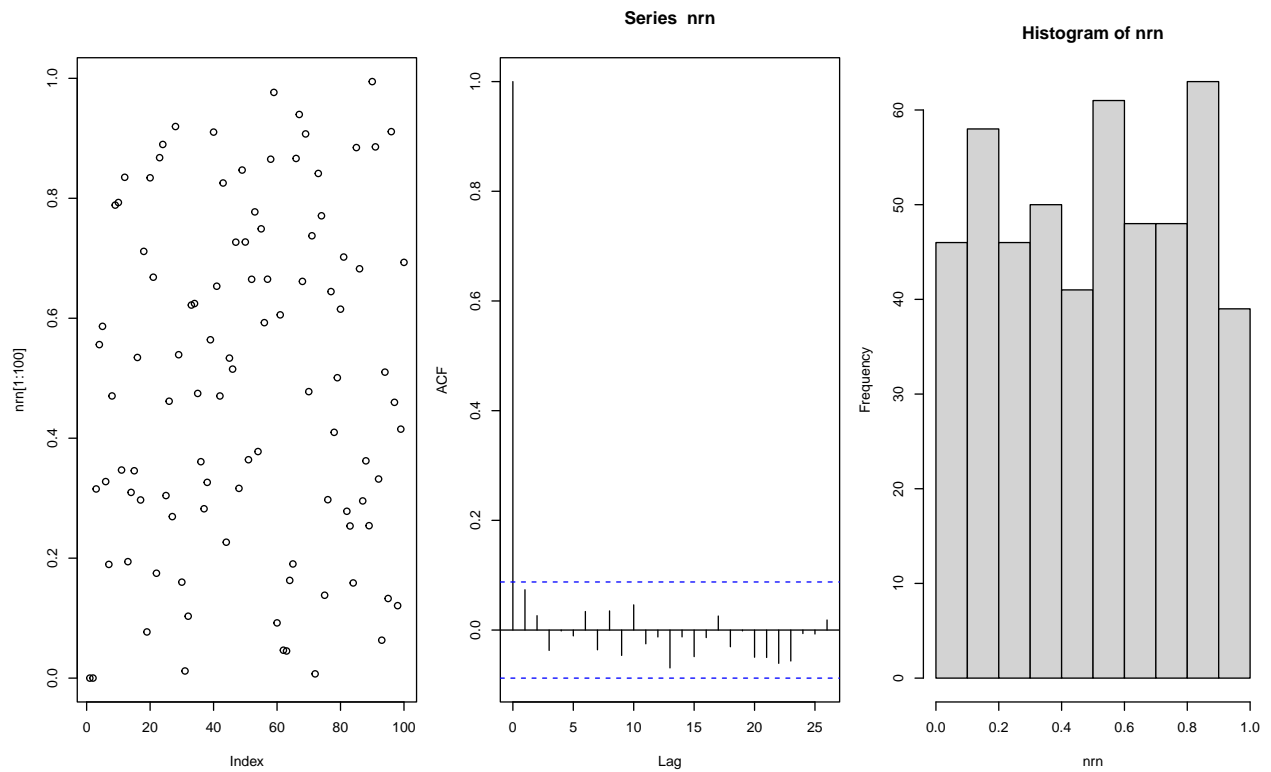
1	Pseudo random numbers	1
2	Inverting CDF	3
3	A Special Transformation for Generating Normal Sample	3
4	Demonstration of CLT and LLN	5
5	An Example of Monte Carlo for Estimating $\pi$	5

## 1 Pseudo random numbers

```
A <- 7^5
M <- 2^31-1

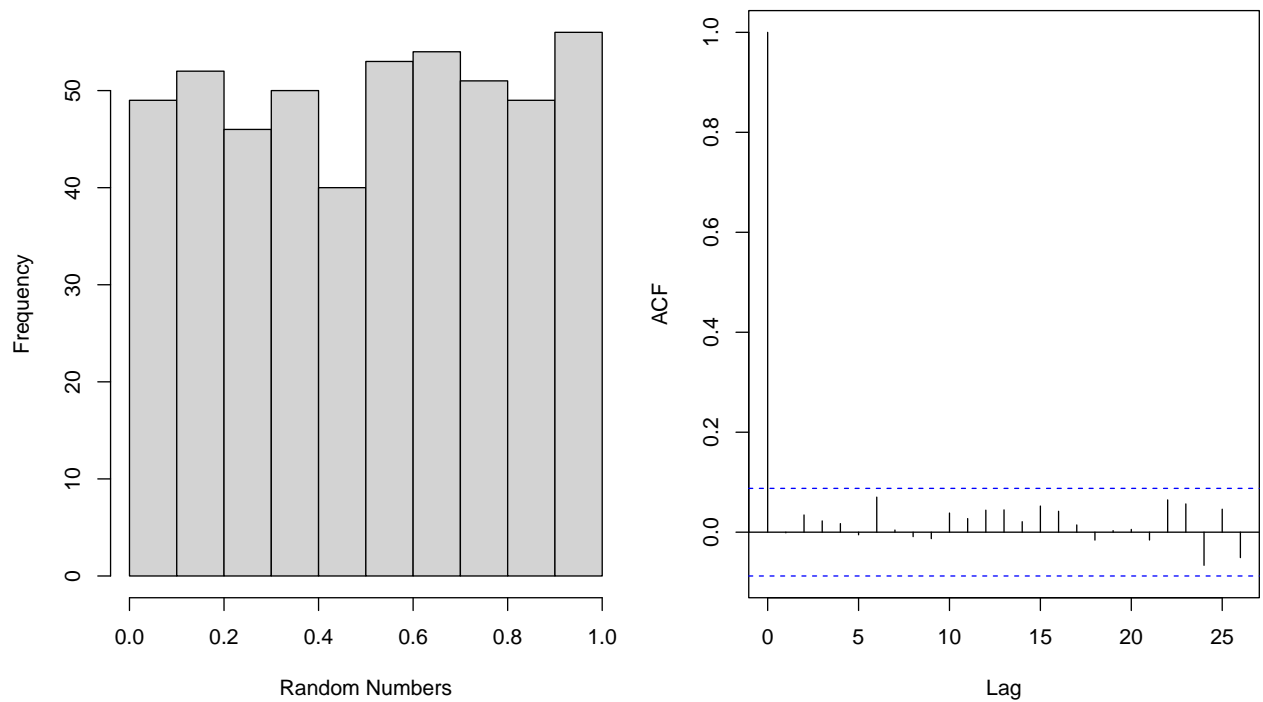
N <- 500
rn <- rep (0, N)
rn[1] <- 10
for (i in 2:length (rn))
{
  rn[i] <- (A * rn[i-1] ) %% M
}

nrn <- rn/(M-1)
par(mfrow=c(1,3),mar=c(4,4,3,1))
plot (nrn[1:100])
acf (nrn)
hist (nrn)
```



```
n <- 500
a <- runif(n)
par(mfrow=c(1,2),mar=c(4,4,3,1))
hist(a,xlab="Random Numbers",main="")

acf(a,main="")
```



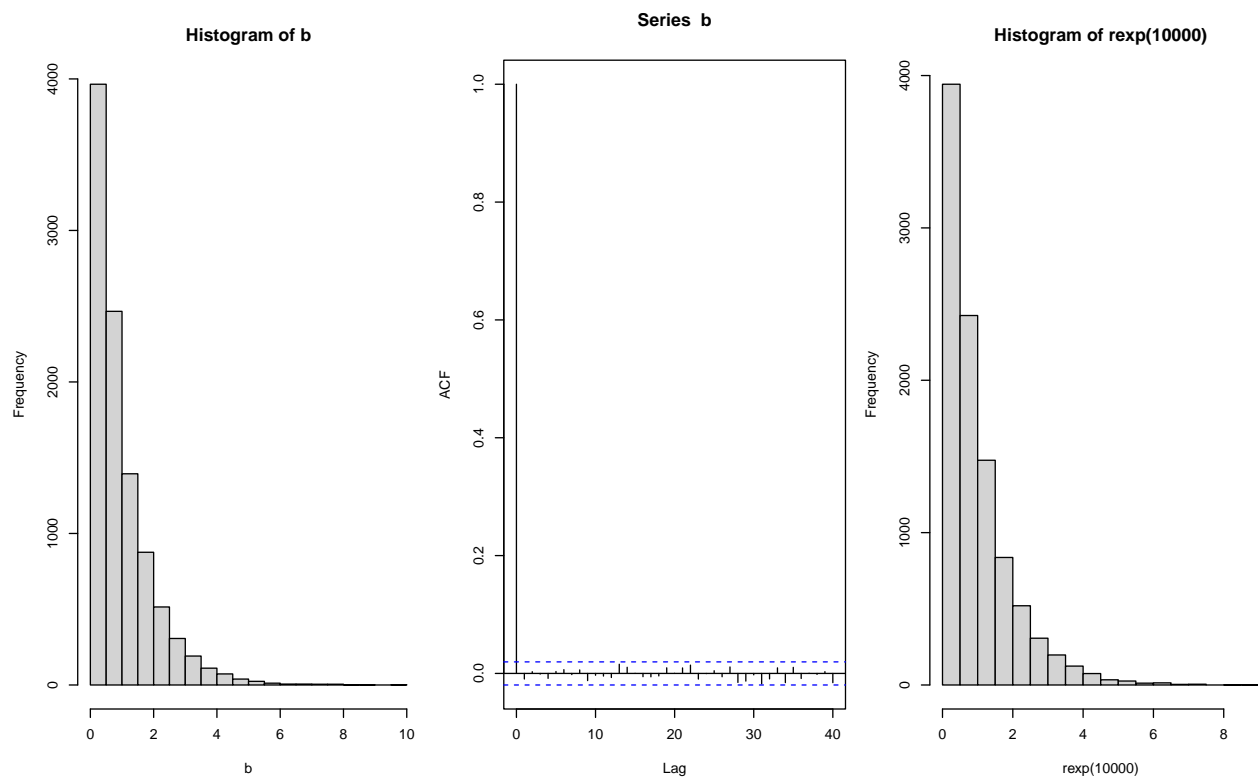
## 2 Inverting CDF

```
# generate exponential random numbers

#use method of inverse cdf to generate iid sample from exp(1)
gen_exp <- function(n)
{
  #generate unif(0,1) random numbers
  u <- runif(n)
  #transform the random numbers
  -log(1-u)
}

b <- gen_exp (10000)
par(mfrow=c(1,3),mar=c(4,4,3,1))
hist (b)
acf (b)

## r built in generators
hist (rexp (10000))
```



## 3 A Special Transformation for Generating Normal Sample

```
gen_normal <- function(n)
{
  #calculates size of random samples, which is greater than half of n
  size_sample <- ceiling(n/2)
```

```

R <- sqrt(2*rexp(size_sample))
theta <- runif(size_sample,0,2*pi)

X <- R*cos(theta)
Y <- R*sin(theta)

c(X,Y)[1:n]
}

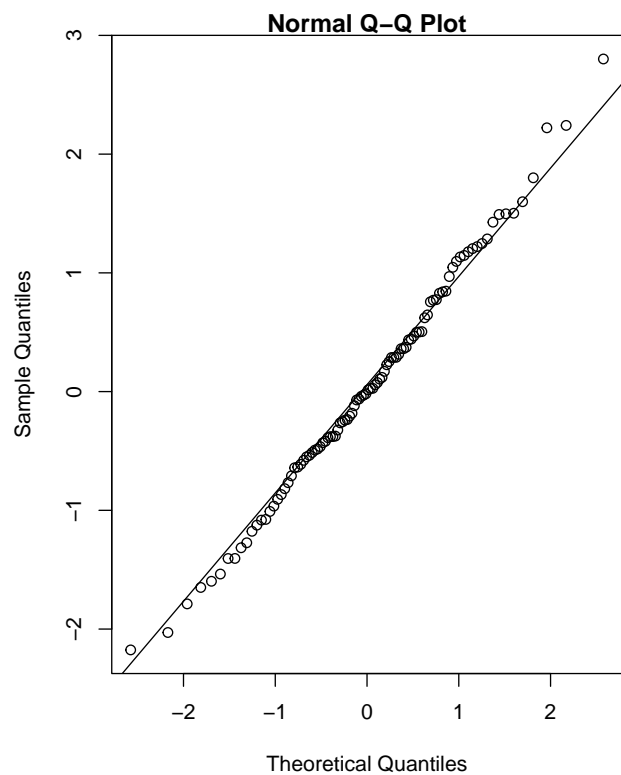
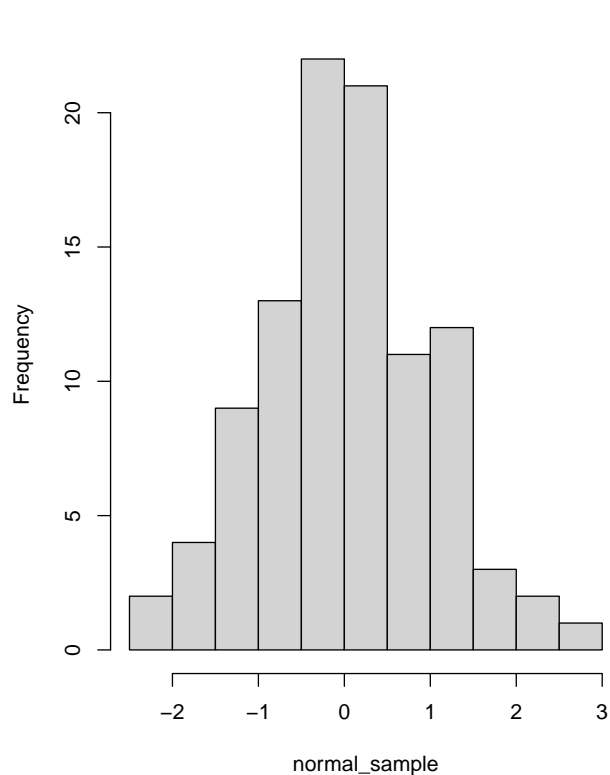
normal_sample <- gen_normal(100)

par(mfrow=c(1,2),mar=c(4,4,1,1))

hist(normal_sample,main="")

qqnorm(normal_sample)
qqline(normal_sample)

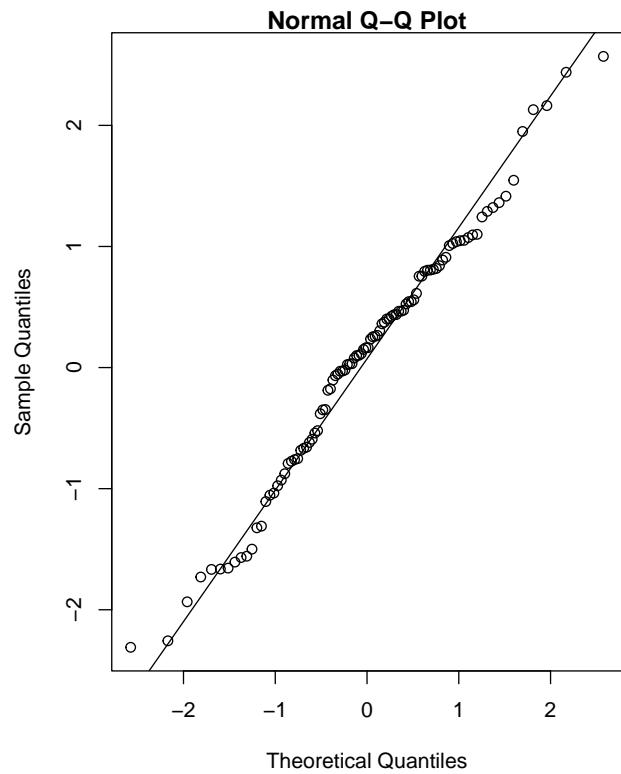
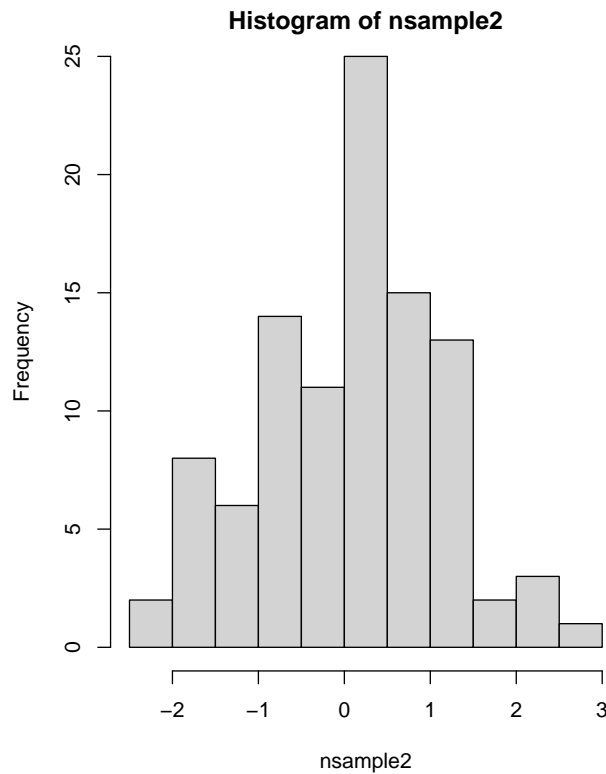
```



```

nsample2 <- rnorm (100)
hist (nsample2)
qqnorm (nsample2)
qqline(nsample2)

```



## 4 Demonstration of CLT and LLN

## 5 An Example of Monte Carlo for Estimating $\pi$

```
#### an application of monte carlo method in estimating pi

# n is the number of samples drawn uniformly from the rectangle (-1,1) * (-1,1)
# an estimate of pi is returned
pi_est_mc <- function(n)
{
  #X and Y are independent, each with marginal distribution unif(-1,1)
  X <- runif(n,-1,1)
  Y <- runif(n,-1,1)

  Z <- 4 * (X^2 + Y^2 <= 1)
  mu <- mean (Z)
  error <- 1.96 * sd (Z) /sqrt (n)
  list (pi.est = mu, error.95perc = error, ci.95perc = mu + c(-error, error))
}

pi_est_mc (100)

## $pi.est
## [1] 3.4
##
## $error.95perc
## [1] 0.2813543
##
```

```
## $ci.95perc
## [1] 3.118646 3.681354
```

```
pi_est_mc (10000)
```

```
## $pi.est
## [1] 3.112
##
## $error.95perc
## [1] 0.03258398
##
## $ci.95perc
## [1] 3.079416 3.144584
```

```
pi_est_mc (100000)
```

```
## $pi.est
## [1] 3.14024
##
## $error.95perc
## [1] 0.01018423
##
## $ci.95perc
## [1] 3.130056 3.150424
```

```
pi_est_mc (1000000)
```

```
## $pi.est
## [1] 3.140321
##
## $error.95perc
## [1] 0.001018383
##
## $ci.95perc
## [1] 3.139302 3.141339
```