

Approximation of Monte Carlo Method

5361 Homework 2

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09/04/2018

Abstract

Consider approximation of standard normal distribution by the Monte Carlo methods.

1 Exercise 1.2

1.1 Functions

I'm going to compare standard Normal distribution:

$$\Phi(t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} e^{-y^2/2} dy$$

with the Monte Carlo methods, where X_i 's are iid $N(0, 1)$ variables:

$$\hat{\Phi}(t) = \frac{1}{n} \sum_{i=1}^n I(X_i \leq t)$$

, when n is large and goes larger.

Experiment with the approximation at $n \in \{10^2, 10^3, 10^4\}$ at $t \in \{0.0, 0.67, 0.84, 1.28, 1.65, 2.32, 2.58, 3.09, 3.72\}$

```
library(knitr)
library(kableExtra)

t <- c(0.0, 0.67, 0.84, 1.28, 1.65, 2.32, 2.58, 3.09, 3.72)
a <- array(data=NA, dim=length(t))

for (i in 1:length(t)){
  a[i] <- pnorm(t[i], mean=0, sd=1)
}

t <- c("t=", t)
a <- c("F(x)=", a)

table1 <- array(c(t, a), dim = c(10, 2))

set.seed(20180913)
norm100 <- rnorm(100)
norm1000 <- rnorm(1000)
norm10000 <- rnorm(10000)

t <- c(0.0, 0.67, 0.84, 1.28, 1.65, 2.32, 2.58, 3.09, 3.72)
b <- array(data=NA, dim=length(t))
c <- array(data=NA, dim=length(t))
```

Table 1: The CDF of Standard Normal Distribution

| t= | F(x)= |
|-----------|-------------------|
| 0 | 0.5 |
| 0.67 | 0.74857110490469 |
| 0.84 | 0.79954580673955 |
| 1.28 | 0.899727432045558 |
| 1.65 | 0.950528531966352 |
| 2.32 | 0.98982956133128 |
| 2.58 | 0.995059984242229 |
| 3.09 | 0.998999217523386 |
| 3.72 | 0.999900388611024 |

Table 2: The Summation of Monte Carlo Methods

| t= | Sum= (n=100) | Sum= (n=1000) | Sum= (n=10000) |
|-----------|---------------------|----------------------|-----------------------|
| 0 | 0.61 | 0.512 | 0.5016 |
| 0.67 | 0.8 | 0.759 | 0.7498 |
| 0.84 | 0.84 | 0.805 | 0.7993 |
| 1.28 | 0.93 | 0.899 | 0.8974 |
| 1.65 | 0.97 | 0.95 | 0.9492 |
| 2.32 | 0.98 | 0.99 | 0.9889 |
| 2.58 | 0.99 | 0.995 | 0.9933 |
| 3.09 | 1 | 0.999 | 0.9988 |
| 3.72 | 1 | 1 | 0.9997 |

```
d <- array(data=NA, dim=length(t))

for (i in 1:length(t)){
  b[i] <- sum(norm100 <= t[i])/100
  c[i] <- sum(norm1000 <= t[i])/1000
  d[i] <- sum(norm10000 <= t[i])/10000
}

t <- c("t=", t)
b <- c("Sum= (n=100)", b)
c <- c("Sum= (n=1000)", c)
d <- c("Sum= (n=10000)", d)

table2 <- array(c(t, b, c, d), dim = c(10, 4))

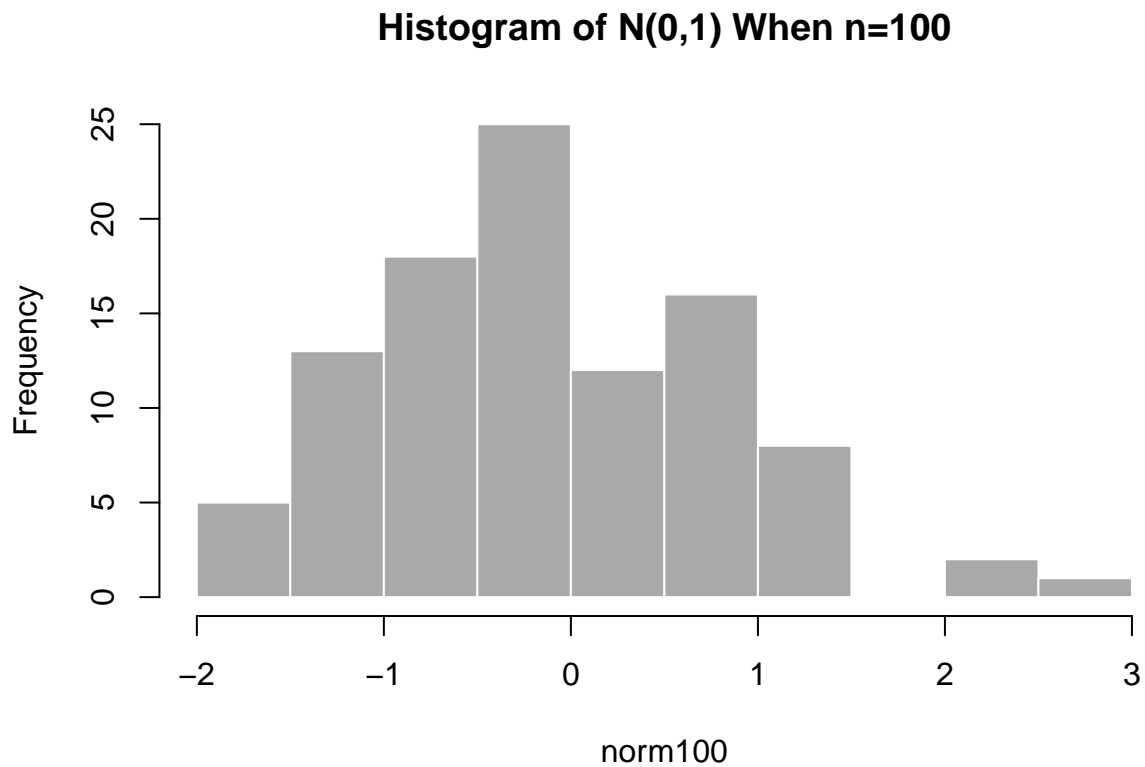
kable(table1, caption = 'The CDF of Standard Normal Distribution', booktabs=T) %>%
  row_spec(1, bold = T, hline_after = T)

kable(table2, caption = 'The Summation of Monte Carlo Methods', booktabs=T) %>%
  row_spec(1, bold = T, hline_after = T)
```

1.2 Figures

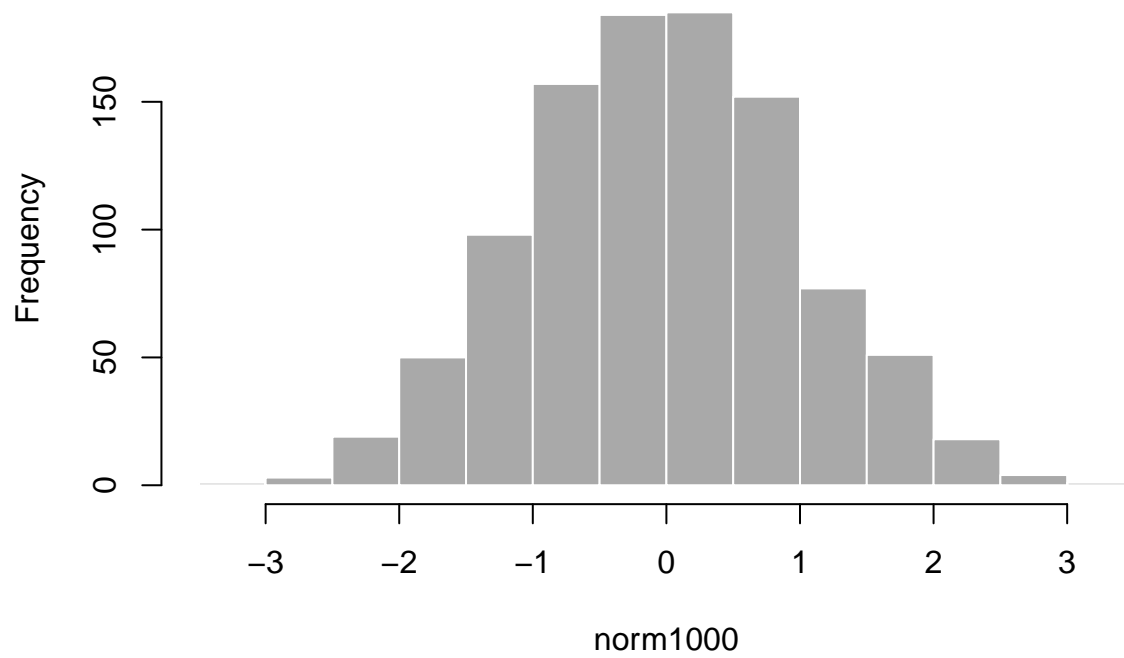
It is visual from graphs of two equations:

```
hist(norm100, col = 'darkgray', border = 'white', main = "Histogram of N(0,1) When n=100")
```



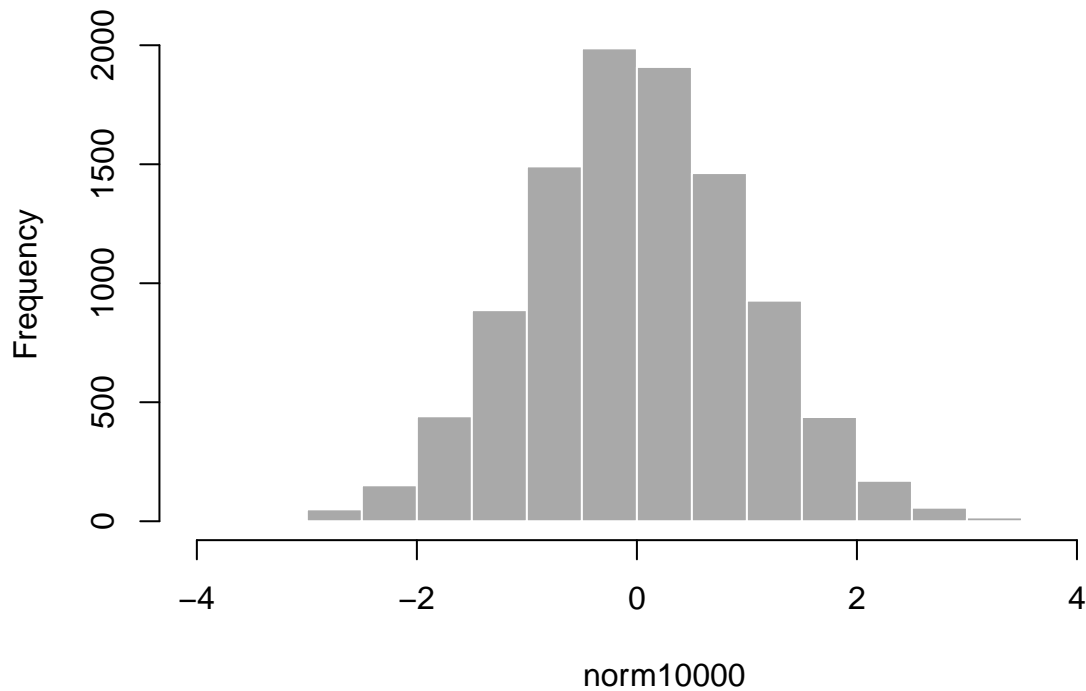
```
hist(norm1000, col = 'darkgray', border = 'white', main = "Histogram of N(0,1) When n=1000")
```

Histogram of $N(0,1)$ When $n=1000$



```
hist(norm10000, col = 'darkgray', border = 'white', main = "Histogram of  $N(0,1)$  When  $n=10000$ ")
```

Histogram of $N(0,1)$ When $n=10000$



Which is obvious that when n increasing, the distribution of samples random chose from standard normal distribution is more close to standard normal distribution, which is a continues distribution.

```
group <- array(NA, dim=11100)
for (x in 1:100){
  group[x] <- "n = 100"
}
for (x in 101:1100){
  group[x] <- "n = 1000"
}
for (x in 1101:11100){
  group[x] <- "n = 10000"
}

norm <- c(norm100, norm1000, norm10000)
boxplot(norm~group, data = NULL,
  main = "Box Plot of N(0,1) with Difference Sample Size",
  xlab = "Sample Size",
  ylab = "Standard Normal Distribution")
```

The box plot displays the distribution of the Standard Normal Distribution for three different sample sizes: $n = 100$, $n = 1000$, and $n = 10000$. The y-axis represents the Standard Normal Distribution, ranging from -4 to 4. For $n = 100$, the distribution is wider, with a median near 0 and several outliers. As the sample size increases to $n = 1000$ and $n = 10000$, the distributions become progressively narrower and more tightly centered around 0, indicating convergence to the standard normal distribution.

Since the highly similarity of two distributions, cdf of $N(0, 1)$ can be simulated by Monte Carlo method in calculation.

The double-precision floating-point format is a computer number format, usually occupying 64 bits in computer memory.

Which is the largest normalized floating-point number. In decimal system, it can be denoted as

[illegible]

```
## [1] 1.797693e+308
```

2.2 .Machine\$double.xmin

Which is the smallest non-vanishing normalized floating-point power of the radix. In decimal system, it can be denoted as

$$(-1)^0(1+1)2^{-1023}$$

In binary system, it can be denoted as 0 0000000001 00₂

```
.Machine$double.xmin
```

```
## [1] 2.225074e-308
```

2.3 .Machine\$double.eps

Which is the smallest positive floating-point number x such that $1 + x \neq 1$. In decimal system, it can be denoted as

2⁻⁵²

In binary system, it can be denoted as 00000000001₂

```
.Machine$double.eps
```

```
## [1] 2.220446e-16
```

2.4 .Machine\$double.neg.eps

Which is a small positive floating-point number x such that $1 - x \neq 1$. In decimal system, it can be denoted as

2-53

In binary system, it can be denoted as 000000000000₂

```
.Machine$double.neg.eps
```

```
## [1] 1.110223e-16
```