

# 2019statcacul

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```
install.packages("foreach") install.packages("tidyverse")
```

#Q1 ##

```
coin <- function(prob=0.3, bet=1, Pe = 6, Pa = 4){
  rec <- tibble(a = Pe, b = Pa)
  while(all(rec>0)){
    u <- runif(1)
    Pe <- ifelse(u<prob, Pe+bet,Pe-bet)
    Pa <- ifelse(u>=prob, Pa+bet,Pa-bet)
    rec <- rec %>% bind_rows(c(a=Pe, b=Pa))
    if(Pe==0 | Pa==0) break()
  }
  rec %>% mutate(idx = 1:n()) %>%
  select(a, b) %>%
  summarise(Gamesu=n(), who_win=ifelse(last(a)==0, "Paul Win!","Peter Win!"), PeterMoney=mean(a))
}

(resault <- coin(prob=0.3, bet=1, Pe=6, Pa=4))
```

```
## # A tibble: 1 x 3
##   Gamesu who_win   PeterMoney
##   <int> <chr>         <dbl>
## 1     11 Paul Win!         3.73
```

0.1

```
rep <- foreach(i=1:1000, .combine = bind_rows) %do%
  {
    coin()
  } %>%
  summarise(Peter_Win_Prob=mean(who_win=="Peter Win!"),Game_Time=mean(Gamesu),Peter_Money_Mean=mean(PeterMoney))

(resault2 <- rep)
```

```
## # A tibble: 1 x 4
##   Peter_Win_Prob Game_Time Peter_Money_Mean
##   <dbl>         <dbl>         <dbl>
## 1     0.028      15.1           3.73
```

#Q2

project 6.B correlation : pearson corr & spearman H0 :  $\rho=0$  이 참일 때  $\rho$ .hat들의 분포를 찾음

$Corr(X_1, \dots, X_n, \text{sample}(c(Y_1, \dots, Y_n)))$

```
library(mvtnorm)
```

```
n<-20 m<-1000 rho0 <- 0 rho <- c(seq(-1,1,length=40)) #alternative M <- length(rho) power1 <- power2 <- power3 <-
```

```
rhotest <- cor.test(samp[,1], samp[,2], method="pearson", rhotestp.value) <- cor.test(samp[, 1], samp[, 2], method =
kendal, rhotestp.value)
```

```
plot(rho, power1, type="l") lines(rho, power2, lty=1) lines(rho, power3, lty=2)
```

```
#Q3
```

0.2

```
m<-10000
mc <- function(m){
  x <-rcauchy(m)
  fg <- as.numeric(x>2)
  a1 <- mean(fg)
  return(a1)
}
mc(m)
```

```
## [1] 0.1547
```

```
##important sampling
```

```
mc.impo1 <- function(m){
  x<-2/runif(m)
  fg <- 1/(2*(pi*(1+1/x^2)))
  b1 <- mean(fg)
  return(b1)
}
mc.impo1(m)
```

```
## [1] 0.1475518
```

0.2.1 important sampling2

```
mc.impo2 <- function(m){
  x<-runif(m, 0, 0.5)
  fg <- dcauchy(x)/2
  b2<-mean(fg)
  return(b2)
}
mc.impo2(m)
```

```
## [1] 0.1476843
```

```
##hit or miss
```

```
mc.hit <- function(m){
  x <- runif(m, 0, 2)
  y <- runif(m, min=0, max=dcauchy(0))
  est <- 1/2 - sum(as.numeric(y<=dcauchy(x)))/m*(2*dcauchy(0)) #2*dcauchy(0) ,sum(as.numeric(y<=dcauchy(x)))/m
  est
}
mc.hit(m)
```

```
## [1] 0.1461031
```

```
##control var
```

```
mccontrol <- function(m){
  x <- runif(m, 0,0.5)
  g <- dcauchy(x)/2
```

```

f1 <- x^2
f2 <- x^4
muf1 <- 1/12
muf2 <- 1/80
L <- lm(g~f1+f2)
#est <- mean(g) - L$coeff[2]*(mean(f1)-muf2) - L$coeff[3]*(mean(f2)-muf2)
est2 <- sum(lm(g ~ f1+f2)$coeff*c(1, muf1, muf2))
est2
}
mccontrol(m)

```

```
## [1] 0.1475834
```

```
##var 비교
```

```

set.seed(1)
rep11 <- foreach(i=1:100, .combine = bind_rows) %do%
{
  tibble(MC1=mc(m),
    MC21=mc.impo1(m),
    MC22=mc.impo2(m),
    MC3=mc.hit(m),
    MC4=mccontrol(m))
} %>%
  summarise(base_vs_impo1=sd(MC1)^2/sd(MC21)^2,
    base_vs_impo2=sd(MC1)^2/sd(MC22)^2,
    base_vs_hit=sd(MC1)^2/sd(MC3)^2,
    base_vs_conval=sd(MC1)^2/sd(MC4)^2)

```

```
(resault4 <- rep11)
```

```

## # A tibble: 1 x 4
##   base_vs_impo1 base_vs_impo2 base_vs_hit base_vs_conval
##   <dbl>         <dbl>         <dbl>         <dbl>
## 1      1011.         1117.         0.889      122286432.

```