

# Homework 3

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*March 7, 2016*

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1. The cycle repeats 4 numbers continuously.

```
x <- 1
for(i in 1:25)
  x <- c(x, (11 * tail(x, 1)) %% 16)
x
```

```
## [1] 1 11 9 3 1 11 9 3 1 11 9 3 1 11 9 3 1 11 9 3 1 11 9
## [24] 3 1 11
```

2. From the 4 examples, we can see that when the cycle reaches 0, it immediately returns to 12. When a higher number like 1000 is entered, it immediately decreases to 11 or lower.

```
x <- 0
for(i in 1:25)
  x <- c(x, (tail(x, 1) + 12) %% 13)
x
```

```
## [1] 0 12 11 10 9 8 7 6 5 4 3 2 1 0 12 11 10 9 8 7 6 5 4
## [24] 3 2 1
```

```
x <- 50
for(i in 1:25)
  x <- c(x, (tail(x, 1) + 12) %% 13)
x
```

```
## [1] 50 10 9 8 7 6 5 4 3 2 1 0 12 11 10 9 8 7 6 5 4 3 2
## [24] 1 0 12
```

```
x <- 100
for(i in 1:25)
  x <- c(x, (tail(x, 1) + 12) %% 13)
x
```

```
## [1] 100 8 7 6 5 4 3 2 1 0 12 11 10 9 8 7 6
## [18] 5 4 3 2 1 0 12 11 10
```

```
x <- 1000
for(i in 1:25)
  x <- c(x, (tail(x, 1) + 12) %% 13)
x
```

```
## [1] 1000 11 10 9 8 7 6 5 4 3 2 1 0 12
## [15] 11 10 9 8 7 6 5 4 3 2 1 0
```

3. Decreasing the 100,000 number a bit so my computer doesn't fail the calc.

```
x <- 1234567
r <- 1234567 / (2^31 - 1)

for(i in 1:99999) {
  x <- c(x, (16807 * tail(x, 1)) %% (2^31 - 1))
  r <- c(r, tail(x, 1) / (2^31 - 1))
}

tbl <- data.frame(x, r)
chisq.test(tbl)
```

```
##
## Pearson's Chi-squared test
##
## data:  tbl
## X-squared = 1.362e-18, df = 99999, p-value = 1
```

The above is almost certainly wrong from the looks of it, not sure what to change here. Let's try runs. Trying this with a package I found.

```
library(randtests)
runs.test(r)
```

```
##
## Runs Test
##
## data:  r
## statistic = -0.32888, runs = 49949, n1 = 50000, n2 = 50000, n =
## 100000, p-value = 0.7422
## alternative hypothesis: nonrandomness
```

4. .

5.

a. First with inverse-normal:

```
normrandit <- function(){
  U <- runif(1)
  return(qnorm(U))
}

itstats <- function(N){
  x <- numeric(0)
  for(i in 1:N){
    x <- c(x, normrandit())
  }
  return(list(mean=mean(x), sd=sd(x)))
}
```

b. Now with Muller:

```
normrandbm <- function(){
  U <- runif(2)
  x <- ((-2*log(U[1]))^(1/2))*cos(2*pi*U[2])
  y <- ((-2*log(U[1]))^(1/2))*sin(2*pi*U[2])
  return(c(x=x, y=y))
}

bmstats <- function(N){
  x <- numeric(0)
  for(i in 1:N){
    x <- c(x,normrandbm())
  }
  return(list(mean=mean(x), sd=sd(x)))
}
```

c. Finally with accept/reject

```
normrandar <- function(){
  repeat{
    U <- runif(2)
    x <- -log(U[1])
    y <- -log(U[2])
    if(y >= ((x-1)^2)/2){
      break
    }
  }
  if (runif(1) >= .5)
    x <- -1 * x
  return(x)
}

arstats <- function(N){
  x <- numeric()
  for(i in 1:N){
    x <- c(x,normrandar())
  }
  return(list(mean=mean(x), sd=sd(x)))
}
```

d. Let's check the means, sd's, and runtimes

```
library(plyr)
means <- data.frame(it=c(), bm=c(), ar=c())
sds <- data.frame(it=c(), bm=c(), ar=c())
times <- data.frame(it=c(), bm=c(), ar=c())

for (n in c(100, 1000, 10000, 100000)) {
  for (i in 1:10) {
    it <- itstats(n)
    bm <- bmstats(n)
```

```

ar <- arstats(n)
means <- rbind(means, c(it$mean, bm$mean, ar$mean, n))
sds <- rbind(sds, c(it$sd, bm$sd, ar$sd, n))
times <- rbind(times, c(system.time(itstats(n))[[3]], system.time(itstats(n))[[3]], system.time(itstats(n))[[3]])
}
}

```

```

colnames(means) <- c("it", "bm", "ar", "n")
colnames(sds) <- c("it", "bm", "ar", "n")
colnames(times) <- c("it", "bm", "ar", "n")

```

```

library(plyr)
means <- read.csv("means.csv")
sds <- read.csv("sds.csv")
times <- read.csv("times.csv")
mean_avgs <- ddply(means, ~n, summarise, mean_it = mean(it), mean_bm = mean(bm), mean_ar = mean(ar));mean_avgs

```

```

##           n      mean_it      mean_bm      mean_ar
## 1 1e+02 -0.0289959302  0.0209949281 -0.0196609236
## 2 1e+03 -0.0169933742 -0.0094790194  0.0059108701
## 3 1e+04  0.0007567270 -0.0009539500 -0.0033564852
## 4 1e+05  0.0005063963 -0.0001194233 -0.0005487528

```

```

sd_avgs <- ddply(sds, ~n, summarise, mean_it = mean(it), mean_bm = mean(bm), mean_ar = mean(ar));sd_avgs

```

```

##           n  mean_it  mean_bm  mean_ar
## 1 1e+02 1.0158512 1.0057870 1.0240378
## 2 1e+03 0.9969124 0.9941471 0.9948150
## 3 1e+04 0.9987530 0.9996241 0.9981487
## 4 1e+05 0.9998424 1.0003970 1.0001452

```

```

times_avgs <- ddply(times, ~n, summarise, mean_it = mean(it), mean_bm = mean(bm), mean_ar = mean(ar));times_avgs

```

```

##           n mean_it mean_bm mean_ar
## 1 1e+02   0.001   0.001   0.000
## 2 1e+03   0.005   0.006   0.006
## 3 1e+04   0.169   0.171   0.142
## 4 1e+05  13.246  13.232  13.260

```

6.

- a. • b. We don't really need to use the method of 1 or 0 to estimate pi, it should estimate fine without checking, and will not give a "4" when a lower N value is given

```

estimatepi <- function(N){
  x <- runif(N)
  y <- runif(N)
  d <- sqrt(x^2 + y^2)
  pi <- (4 * sum(d < 1.0) / N)
  se <- sd(d) / sqrt(length(d))
}

```

```

ci <- qnorm(0.95)*sd(d)/sqrt(N)
return(list(pi=pi, se=se, ci=ci))
}

```

c. Let's check how large N needs to be for the estimate to be accurate.

```

values <- data.frame(pi=c(),se=c(),ci=c())
x <- seq(1000,10000,by=500)
for (n in x) {
  values <- rbind(values, estimatepi(n))
}
rownames(values) <- seq(1000,10000,by=500)

values$lower <- 3.14 - values$ci
values$upper <- 3.14 + values$ci
values$interval <- values$upper - values$lower
values

```

##	pi	se	ci	lower	upper	interval
## 1000	3.076000	0.008764781	0.014416782	3.125583	3.154417	0.028833564
## 1500	3.096000	0.007496967	0.012331414	3.127669	3.152331	0.024662828
## 2000	3.186000	0.006277706	0.010325908	3.129674	3.150326	0.020651816
## 2500	3.148800	0.005708940	0.009390371	3.130610	3.149390	0.018780743
## 3000	3.137333	0.005195896	0.008546488	3.131454	3.148546	0.017092977
## 3500	3.192000	0.004811151	0.007913639	3.132086	3.147914	0.015827278
## 4000	3.156000	0.004481106	0.007370764	3.132629	3.147371	0.014741528
## 4500	3.148444	0.004263198	0.007012337	3.132988	3.147012	0.014024674
## 5000	3.156800	0.004032255	0.006632469	3.133368	3.146632	0.013264939
## 5500	3.139636	0.003845109	0.006324641	3.133675	3.146325	0.012649282
## 6000	3.146000	0.003667488	0.006032482	3.133968	3.146032	0.012064963
## 6500	3.179077	0.003471730	0.005710487	3.134290	3.145710	0.011420975
## 7000	3.125143	0.003396492	0.005586731	3.134413	3.145587	0.011173463
## 7500	3.122667	0.003319845	0.005460659	3.134539	3.145461	0.010921317
## 8000	3.148500	0.003186021	0.005240539	3.134759	3.145241	0.010481078
## 8500	3.116235	0.003121408	0.005134259	3.134866	3.145134	0.010268518
## 9000	3.145333	0.002977020	0.004896762	3.135103	3.144897	0.009793524
## 9500	3.156211	0.002918253	0.004800099	3.135200	3.144800	0.009600197
## 10000	3.128800	0.002873611	0.004726670	3.135273	3.144727	0.009453340