# STA 3431 Assignment #1

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## Question 1

In this question, we choose another set of parameters for LCG. As discussed in class, by the Hull–Dobell Theorem, we want to make sure:

```
    gcd(b, m) = 1
    every "prime or 4" divisor of m also divides a - 1.
```

```
# define random function
m = 2^32
a = 4 * 69069 + 1
b = 23606797 * 7
latestval = 12345
nextrand = function() {
    latestval <<- (a * latestval + b)%m
    return(latestval/m)
}

# record 100000 random variables
rand_vals = c(1e+06)
for (i in 1:1e+06) {
    rand_vals[i] = nextrand()
}</pre>
```

As we choose m to be 2^32, we only need b to be odd and a - 1 to be a multiple of 4. To avoid similarities between Un and Un-1, we choose a relatively large a.

Let's see the statistics compared to theoretical values of Uniform(0, 1).

```
# compare statistics with theoretical values
limits= c(100, 1000, 10000, 100000)
for (lmt in limits) {
   cat("For the first ", lmt, "observations:\n")
   cat("Real mean: ", mean(rand_vals[1:lmt]), ", should be: ", 0.5, '\n')
   cat("Real standard deviation: ", sd(rand_vals[1:lmt]), ", should be: ", sqrt(1/12), '\n')
}

## For the first 100 observations:
## Real mean: 0.4692917 , should be: 0.5
## Real standard deviation: 0.3109622 , should be: 0.2886751
## For the first 1000 observations:
## Real mean: 0.5035736 , should be: 0.5
## Real standard deviation: 0.2943177 , should be: 0.2886751
## For the first 10000 observations:
## Real mean: 0.4992396 , should be: 0.5
```

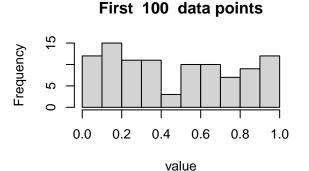
```
## Real standard deviation: 0.2905168 , should be: 0.2886751
## For the first 1e+05 observations:
## Real mean: 0.5003146 , should be: 0.5
## Real standard deviation: 0.2885985 , should be: 0.2886751
```

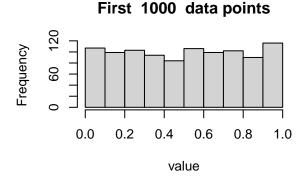
#### Randomness

#### Uniformity

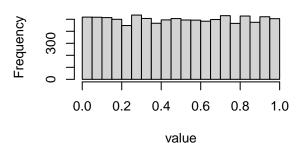
Let's see the distribution of the numbers generated by this algorithm.

```
# plot frequency of the first N variables.
par(mfrow = c(2, 2))
for (lmt in limits) {
    hist(rand_vals[1:lmt], main = paste("First ", lmt, " data points"), xlab = "value")
}
```

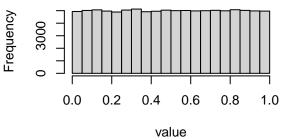








## First 1e+05 data points



From the histograms alone, we can see that, as the number of observations increases, the values are evenly (uniformly) distributed between 0 and 1. To quantify this, we perform a Chi squared test

### Question 2

As discussed in class, we can simulate exponential distribution and normal distribution using uniform distribution.

```
# simulation size
n = 1000

# function for acquiring n Uniform random numbers
get_randvals <- function(n) {
  rand_vals = c(n)
  for (i in 1:n) {</pre>
```

```
rand_vals[i] = nextrand()
 }
 return(rand_vals)
}
# function for estimating required expectation
get_exp <- function() {</pre>
  \# lambda = 3
 U1 = get_randvals(n)
 Y = -log(U1) / 3
  # use Box-Muller transformation
 U2 = get_randvals(n)
 U3 = get_randvals(n)
  Z = sqrt(2 * log(1/U2)) * cos(2 * 3.14159265 * U3)
  # get vars
 X = abs((Y^2) * (Z^5) * sin((Y^3) * (Z^2)))
  # compute and output the mean and standard error
 m = mean(X)
  se = sd(X) / sqrt(n)
  cat("MC: ", m, " +- ", se, " (n=", n, ")", "\n", sep='')
  cat(" 95% C.I.: (", m-1.96*se, ",", m+1.96*se, ")\n", sep='')
n = 1000000
get_exp()
## MC: 0.8465307 +- 0.01100011 (n=1e+06)
   95% C.I.: (0.8249705,0.8680909)
get_exp()
## MC: 0.8255937 +- 0.01109805 (n=1e+06)
   95% C.I.: (0.8038416,0.8473459)
get_exp()
## MC: 0.8367128 +- 0.01068597 (n=1e+06)
   95% C.I.: (0.8157683,0.8576573)
We can see that the variance is still pretty high. Try increasing the simulation size to see if it gets better.
# Incrase n to reduce standard error
n = 10000000
get_exp()
## MC: 0.8422648 +- 0.00378406 (n=1e+07)
   95% C.I.: (0.834848,0.8496815)
get_exp()
## MC: 0.8319338 +- 0.003634367 (n=1e+07)
## 95% C.I.: (0.8248105,0.8390572)
get_exp()
```

```
## MC: 0.8356876 +- 0.003617307 (n=1e+07)
## 95% C.I.: (0.8285977,0.8427775)
```

The estimated expectation value is 0.83 with the approximate 95% confidence interval being about (0.82, 0.84). This estimation is not very accurate as the estimated standard error is relatively high ( $\sim 0.0036$ ) even though we are using a large n (10000000). This is because the term  $|Y^2Z^5|$  has high variance when  $Y \sim \text{Exponential}(3)$  and  $Z \sim \text{Normal}(0,1)$ . In this case, even with large simulation size, the result of "classical" Monte Carlo method still fluctuates a lot between different runs.

## Question 3

My student number is 1003118547. Using the last 4 digits, we have:

$$A = 8, B = 5, C = 4, D = 7$$