

# Homework 3

## DATA604 Simulation and Modeling

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```
##
## Attaching package: 'gplots'

## The following object is masked from 'package:stats':
##
##      lowess

library(knitr)
library(knitr::knitcitations)
library(RefManager)

cleanbib()

cite_options(style="markdown")

bibPkgTseries <- bibentry(bibtype="Misc",
  author=personList(person(family="Trapletti", given="Adrian"),
    person(family="Hornik", given="Kurt")),
  journal="Annual Review of Sociology",
  title="tseries: Time Series Analysis and Computational Finance",
  year=2015,
  note="R package version 0.10-34",
  url="http://CRAN.R-project.org/package=tseries")
```

1

Starting with  $X_0 = 1$ , write down the entire cycle for  $X_i = 11X_{i-1} \bmod(16)$

```
fn1 <- function(x0)
{
  df <- data.frame(X=c(), R=c())
  x <- x0
  continue <- TRUE

  while(continue)
  {
    xi <- (11 * x) %% 16
    df <- rbind(df, data.frame(X=x, R=xi))
    x <- xi

    if(xi == x0)
    {
      break
    }
  }

  return(df)
}

res <- fn1(1)
```

X	R
1	11
11	9
9	3
3	1

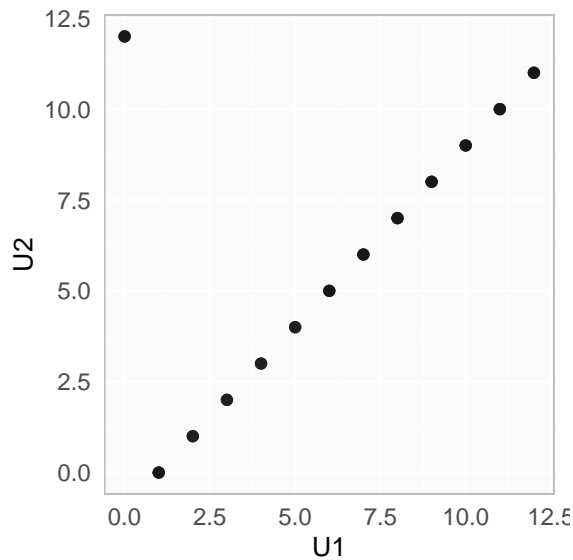
## 2

\*Using the LCG provided below:  $X_i = (X_{i-1} + 12) \bmod(13)$ , plot the pairs  $(U_1, U_2), (U_2, U_3), \dots$  and observe the lattice structure obtained. Discuss what you observe.

```
fn2 <- function(x0, max=100)
{
  df <- data.frame(U1=c(), U2=c())
  x <- x0

  for(i in 1:max)
  {
    xi <- (x + 12) %% 13
    df <- rbind(df, data.frame(U1=x, U2=xi))
    x <- xi
  }

  return(df)
}
# Call the function starting at x0=1
res <- fn2(1)
```



The chart above suggests there are only 13 points, but actually the LCG cycle period is 13 and numbers are repeating.

U1	U2
1	0
0	12
12	11
11	10
10	9
9	8
8	7
7	6

U1	U2
6	5
5	4
4	3
3	2
2	1
1	0

### 3

Implement the pseudo-random number generator:

$$X_i = 16807X_{i-1} \bmod(2^{31} - 1)$$

Using the seed 1234567, run the generator for 100,000 observations. Perform a chi-square goodness-of-fit test on the resulting PRN's. Use 20 equal-probability intervals and level  $\alpha = 0.05$ . Now perform a runs up-and-down test with  $\alpha = 0.05$  on the observations to see if they are independent.

```
fnLCG3 <- function(seed = 1, n = 1)
{
  rands <- rep(NA, n)
  x <- seed
  modVal <- (2^31 - 1)

  for(i in 1:n)
  {
    xi <- (16807 * x) %% (modVal)
    rands[i] <- xi
    x <- xi
  }

  return(rands)
}

n=100000
rn <- fnLCG3(1234567, n)
```

1422014746
456328559
849987676
681650688
1825340118
1687465831

### Chi-Square Test

```
intervals <- 20
maxRn <- max(rn)
minRn <- min(rn)
intWidth <- (maxRn - minRn) / intervals
lwr <- minRn
dfCounts <- data.frame(intID=c(), count=c())
# Bin the data ourselves, I'd guess there
# is an easier way, but this will do.
```

```

for(i in 1:intervals)
{
  upr <- lwr + intWidth
  inRange <- rn[lwr <= rn & rn < upr]
  dfCounts <- rbind(dfCounts, data.frame(intID=i, count=length(inRange)))
  # setup for next interval range
  lwr <- upr
}
# Do our own Chi-Squared test
Expected <- (100000 / intervals)
chi2 <- sum((dfCounts$count - Expected)^2 / Expected)
chi2

```

```
## [1] 14.7762
```

```

# Use built-in function to compare
chiTest <- chisq.test(dfCounts$count)
chiTest

```

```

##
## Chi-squared test for given probabilities
##
## data: dfCounts$count
## X-squared = 14.776, df = 19, p-value = 0.7367

```

The p-value = 0.7367029 is not less than  $\alpha = 0.05$ , therefore we doesn't reject the null hypothesis that the distrubtion is uniform.

intID	count
1	5069
2	5028
3	5044
4	5087
5	4948
6	4953
7	4937
8	4933
9	4900
10	4957
11	5088
12	4994
13	5076
14	5019
15	5002
16	5067
17	4981
18	4914
19	5062
20	4940

## Runs Up-and-Down Test

Using the `tseries` package, we execute the Runs test (Trapletti and Hornik, 2015). First we have to construct the +/- vector. Here we simply convert to boolean.

```

s <- rep(NA, n - 1)
for(i in 1:n - 1)
{
  s[i] <- rn[i] < rn[i + 1]
}

runsTest <- runs.test(as.factor(s))
runsTest

```

```

##
## Runs Test
##
## data:  as.factor(s)
## Standard Normal = 105.84, p-value < 2.2e-16
## alternative hypothesis: two.sided

```

Based on the p-value < 0.05, we reject the null hypothesis and conclude there is not evidence to support independence in the psuedo-random numbers.

#### 4.

*Give inverse-transforms, composition, and acceptance-rejectino algorithms for generating fromt he following density:*

$$f(x) = \begin{cases} \frac{3x^2}{2} & -1 \leq x \leq 1 \\ 0 & otherwise \end{cases}$$

## References

Trapletti, A. and K. Hornik. tseries: Time Series Analysis and Computational Finance. R package version 0.10-34. 2015. URL: <http://CRAN.R-project.org/package=tseries>.