

Why Random Sequences?

istical samples

gorithms

y:

random keys

g streams of random bits (e.g., SSL xor's your data with a pseudo-random bit stream that only you and the can generate).

se, games

3:54 2018

CS61B: Lecture #32 2

Pseudo-Random Sequences

table, a "truly" random sequence is difficult for a com-
man) to produce.

poses, need only a sequence that satisfies certain sta-
tistics, even if deterministic.

e.g., cryptography) need sequence that is *hard* or *im-*
predict.

om sequence: deterministic sequence that passes some
statistical tests.

, look at lengths of *runs*: increasing or decreasing con-
sequences.

ly, statistical criteria to be used are quite involved. For
Knuth.

3:54 2018

CS61B: Lecture #32 4

What Can Go Wrong (I)?

is, many impossible values: E.g., a , c , m even.

erns. E.g., just using lower 3 bits of X_i in Java's 48-bit
to get integers in range 0 to 7. By properties of modular

$$\begin{aligned} \text{mod } 8 &= (25214903917X_{i-1} + 11 \bmod 2^{48}) \bmod 8 \\ &= (5(X_{i-1} \bmod 8) + 3) \bmod 8 \end{aligned}$$

period of 8 on this generator; sequences like

0, 1, 3, 7, 1, 2, 7, 1, 4, ...

le. This is why Java doesn't give you the raw 48 bits.

3:54 2018

CS61B: Lecture #32 6

CS61B Lecture #32

om Numbers (Chapter 11)

e random sequences?

andom sequences"?

om sequences.

ne.

a library classes and methods.

utations.

3:54 2018

CS61B: Lecture #32 1

What Is a "Random Sequence"?

"a sequence where all numbers occur with equal fre-

3, 4, ...?

ow about: "an unpredictable sequence where all numbers
qual frequency?"

0, 1, 1, 2, 2, 2, 2, 3, 4, 4, 0, 1, 1, 1, ...?

it is wrong with 0, 0, 0, 0, ... anyway? Can't that occur
election?

3:54 2018

CS61B: Lecture #32 3

Generating Pseudo-Random Sequences

as you might think.

implex jumbling methods can give rise to bad sequences.

quential method is a simple method used by Java:

$$\begin{aligned} X_0 &= \text{arbitrary seed} \\ X_i &= (aX_{i-1} + c) \bmod m, \quad i > 0 \end{aligned}$$

large power of 2.

sults, want $a \equiv 5 \bmod 8$, and a , c , m with no common

erator with a *period of m* (length of sequence before
and reasonable *potency* (measures certain dependencies
ent X_i .)

ts of a to "have no obvious pattern" and pass certain
(see Knuth).

$= 25214903917$, $c = 11$, $m = 2^{48}$, to compute 48-bit
om numbers. It's good enough for many purposes, but
aphically secure.

3:54 2018

CS61B: Lecture #32 5

Additive Generators

erator:

$$X_n = \begin{cases} \text{arbitrary value,} & n < 55 \\ (X_{n-24} + X_{n-55}) \bmod 2^e, & n \geq 55 \end{cases}$$

es than 24 and 55 possible.

period of $2^f(2^{55} - 1)$, for some $f < e$.

mentation with circular buffer:

```
55;
+31) % 55]; // Why +31 (55-24) instead of -24?
/* modulo 232 */
```

. 54] is initialized to some "random" initial seed val-

3:54 2018

CS61B: Lecture #32 8

aphic Pseudo-Random Number Generator Example

a good **block cipher**—an encryption algorithm that en-
s of N bits (not just one byte at a time as for Enigma).
ample.

rovide a key, K , and an initialization value I .

udo-random number is now $E(K, I + j)$, where $E(x, y)$ is
on of message y using key x .

3:54 2018

CS61B: Lecture #32 10

Adjusting Range (II)

bias problems when n does not evenly divide 2^{48} , Java
values after the largest multiple of n that is less than

```
integer in the range 0 .. n-1, n>0. */
t(int n) {
    next random long (0 ≤ X < 248);
    yk for some k)
    n top k bits of X;

    = largest multiple of n that is < 248;
    i >= MAX)
    next random long (0 ≤ X < 248);
    i / (MAX/n);
```

3:54 2018

CS61B: Lecture #32 12

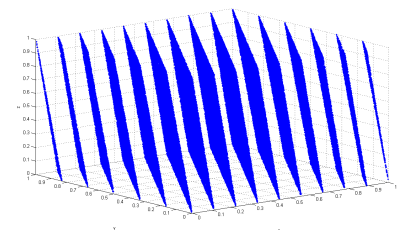
What Can Go Wrong (II)?

ds to bad correlations.

s IBM generator RANDU: $c = 0$, $a = 65539$, $m = 2^{31}$.

U is used to make 3D points: $(X_i/S, X_{i+1}/S, X_{i+2}/S)$,
es to a unit cube, ...

be arranged in parallel planes with voids between. So
ts" won't ever get near many points in the cube:



Luis Sanchez at English Wikipedia - Transferred from en.wikipedia to Commons
, CC BY-SA 3.0, <https://commons.wikimedia.org/w/index.php?curid=3832343>

3:54 2018

CS61B: Lecture #32 7

aphic Pseudo-Random Number Generators

form of linear congruential generators means that one
future values after seeing relatively few outputs.

you want **unpredictable** output (think on-line games in-
y or randomly generated keys for encrypting your web

aphic pseudo-random number generator (CPRNG) has the
hat

ts of a sequence, no polynomial-time algorithm can guess
bit with better than 50% accuracy.

current state of the generator, it is also infeasible to
ct the bits it generated in getting to that state.

3:54 2018

CS61B: Lecture #32 9

Adjusting Range and Distribution

sequence of numbers, X_i , from above methods in range
 2^{48} , how to get uniform random integers in range 0 to

easy: use top k bits of next X_i (bottom k bits not as

be careful of slight biases at the ends. For example, if
 $X_i/(2^{48}/n)$ using all integer division, and if $(2^{48}/n)$ gets
 n , then you can get n as a result (which you don't want).

fix that by computing $(2^{48}/(n-1))$ instead, the proba-
ing $n-1$ will be wrong.

3:54 2018

CS61B: Lecture #32 11

Alternative Selection Algorithm (Floyd)

```
once of K distinct integers
0<=K<=N. */
selectRandomIntegers(int N, int K, Random R)

// IntList();

// K; i < N; i += 1) {
// s in S are < i
// findInt(i+1); // 0 <= s <= i < N
// get(j) for some j
// value i (which can't be there
// after the s (i.e., at a random
// other than the front)
// i);

// random value s at front
// );
```

Example

<i>i</i>	<i>s</i>	<i>S</i>
5	4	[4]
6	2	[2, 4]
7	5	[5, 2, 4]
8	5	[5, 8, 2, 4]
9	4	[5, 8, 2, 4, 9]

selectRandomIntegers(10, 5, R)