

# *RANDOM VARIABLE GENERATION (PART 2)*

And now for something  
completely different...

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# The Dirichlet Distribution

- Now that we've done the Gamma distribution...
  - In reality, we're not interested in the Gamma directly for our project
  - Rather, we are interested 'cause can be used to generate a Dirichlet RV
- Dirichlet
  - Real-vector-valued RV
  - Parameter set is a list of  $m$  real values, each larger than zero
  - Output is a vector-valued list of real numbers from 0 to 1:

$$\langle 0.2, 0.5, 0.2, 0.1 \rangle$$

- Constrained so that they sum to one
- Super important, 'cause used to model a random vector of probabilities

# Turns Out That Generating a Dirichlet is Easy

- Given params  $k_1, k_2, \dots, k_m$
- Generate  $m$  random values, using  $rv_i \sim \text{Gamma}(k_i, c)$
- Then  $i$ th entry in output vector is simply

$$\frac{rv_i}{\sum_{j=1}^m rv_j}$$

- That's it!

# Generating a Multinomial

- MN simulates having a large (infinite) number of balls in a bag
- Are  $m$  colors for the balls
- Proportion of color  $i$  is  $p_i$
- Then you reach in and select  $n$  balls at random
- $i$ th entry in vector is how many of color  $i$  that you selected
- So have a vector of ints no less than zero
- Where L1 norm is  $n$

# How To Implement?

- Discrete, so generally easier than continuous
- Assume  $n = 1$
- Just draw a random value  $rv$  from zero to one
- Compute  $i$  where

$$rv \geq \sum_{j=1}^{i-1} p_j$$

- And

$$rv \leq \sum_{j=1}^i p_j$$

- Then return a vector of all zeros, except for a one at pos  $i$

## What If $n$ exceeds one??

- Just repeat the last process  $n$  times
- And add up the  $n$  resulting vectors
- Simple! But can you do this really fast?
- Sure! Generate  $n$  random values, using  $rv_i \sim \text{Uniform}(0, 1)$
- Sort 'em
- Then make a linear pass through random values and the  $p_i$ 's
- And merge 'em! Will do on the board...
- Running time?  $O(n + m + n \log n) = ??$

# Some Notes on PRNG

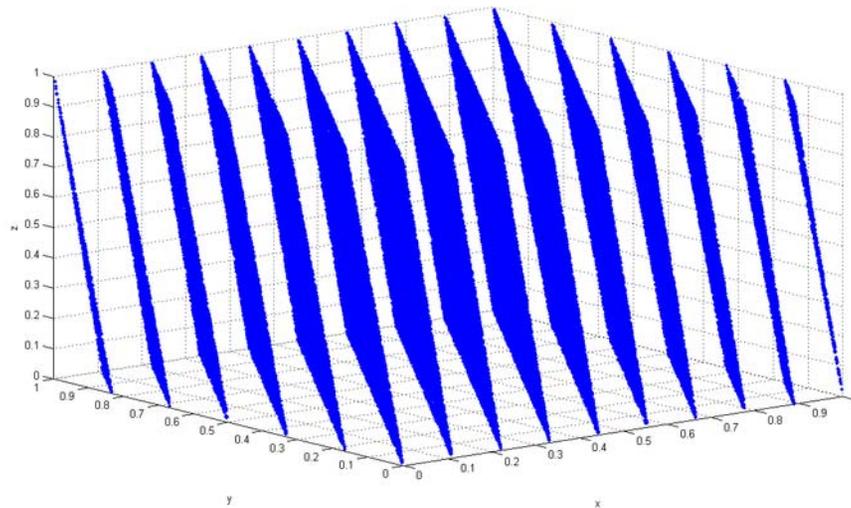
- All of this depends on being able to generate random doubles
  - From zero to one
- Easy if you can generate a random sequence of bits
- But how to generate that sequence of bits? Use a PRNG!
- We're not asking you to implement one...
  - Java comes with an imp. of a classic PRNG: the “linear congruential”
- Will talk briefly about this. Want to learn more?
  - Chapter 7 of classic CS book “Numerical Recipes”
  - [http://books.google.com/books/about/Numerical\\_recipes.html?id=1aAOdzK3FegC](http://books.google.com/books/about/Numerical_recipes.html?id=1aAOdzK3FegC)

# The Linear Congruential Method

- Defined by  $X_{n+1} = (aX_n + c) \bmod m$ 
  - In this formula,  $X_n$  is the  $n$ th pseudo-random number generated
  - $X_0$  is known as the “seed”
  - Note: can always recreate sequence of bits by going back to  $X_0$
- The “period” of a PRNG is the time until it loops back
  - Obviously, we want a period of  $m$  here
  - Theoretically guaranteed if:
    - (a)  $a - 1$  is divisible by all prime factors of  $m$
    - (b) if  $m$  is a multiple of 4, then  $a - 1$  is a multiple of 4
    - (c)  $c$  and  $m$  are relatively prime

# The Linear Congruential Method

- Before people understood this stuff...
  - Were some classically bad parameter choices
- Infamous example was IBM's "RANDU" routine
  - Following picture shamelessly stolen from Wikipedia



- LCM still widely used, though more modern PRNGs available!

# Some Closing Notes on A4

- You are asked to provide several implementations for

`IRandomGenerationAlgorithm <ReturnType>`

- In the end, we felt it was quite challenging...

- So we are giving you part of the design

- In particular, the abstract class interface for our various RNG algorithms:

`ARandomGenerationAlgorithm <ReturnType>`

- Key idea: put the PRNG in the abstract class

- So abstract away problem of generating uniform numbers...

- Totally handled by abstract

- Have two constructors in both abstract and concrete. One that accepts a seed...

- And one that accepts an “IPRNG” object

- IPRNG wraps up the PRNG algorithm; we’ve given you “PRNG” class

## Some Closing Notes on A4

- In concrete, just call “super()” and then set up local structures
- In abstract...
  - If you get the seed, use it to set up a new PRNG object as the default
  - If you get an IPRNG, then use it
- This is also useful for “linking” RV generation algorithms
  - Example: if Dirichlet uses a bunch of Gammas...
  - They should all use Dirichlet’s IPRNG object
  - That way you know they are all using the same PRNG sequence

# Some Closing Notes on A4

- How do A4 test cases work?
  - After all, two “correct” imps may spit out different random values
- We instantiate a random variable, then use it to gen many values
- Then compare theoretical vs. observed mean and std. dev.
- Highly unlikely you can have a bad imp that passes the test case
- Downside? Can be difficult to debug
  - Not always a clear path from bad mean/std. dev back to bug in your code
- Sooo... start early!
  - Might be easy, but also could be quite challenging

Questions?