

From Monte Carlo Simulation to Bootstrap

Eric Slud, Mathematics Department, UMCP

Objective: to explain an "experimental" approach to Probability & Statistics via Simulation

Outline

O. Definition of Probability & Simulation

I. Simulation-based estimation of Probabilities

II. Simulation in relation to Data: Histograms and Densities

III. Resampling from Data: Why Do It ?

What is Probability ?

- A rule for assigning numbers between 0 and 1 to *Events*
- Obeys combination rules the same as Relative Frequencies
 - 1 = certain to occur
 - Prob's add for unions of disjoint events
- Definition of Relative Frequency: if a Random Experiment is replicated a large number N times, independently (*with mutual non-interference*) by same mechanism, and event E occurs n_E times, its **relative frequency** of occurrence is n_E/N .

Probability as Limiting Relative Frequency

Probability axioms are obeyed by relative frequencies.

Formal mathematics definition of Probability as Set-Function plus def'n of **independent identical-mechanism replications**

$$E_1, E_2, \dots, E_N : \begin{cases} P(E_i) \text{ same for all } i, \text{ and for } j\text{'s distinct} \\ P(E_{j_1} \cap E_{j_2} \cap \dots \cap E_{j_k}) = P(E_{j_1}) \dots P(E_{j_k}) \end{cases}$$

leads to mathematical theorem **Law of Large Numbers** saying:

$$\text{as } N \rightarrow \infty, \quad \frac{1}{N} \sum_{j=1}^N I[E_j] \rightarrow P(E_1)$$

We want to implement this computationally !

What is Monte Carlo Simulation ?

Ingredient #1: Dynamical Random Number Generator

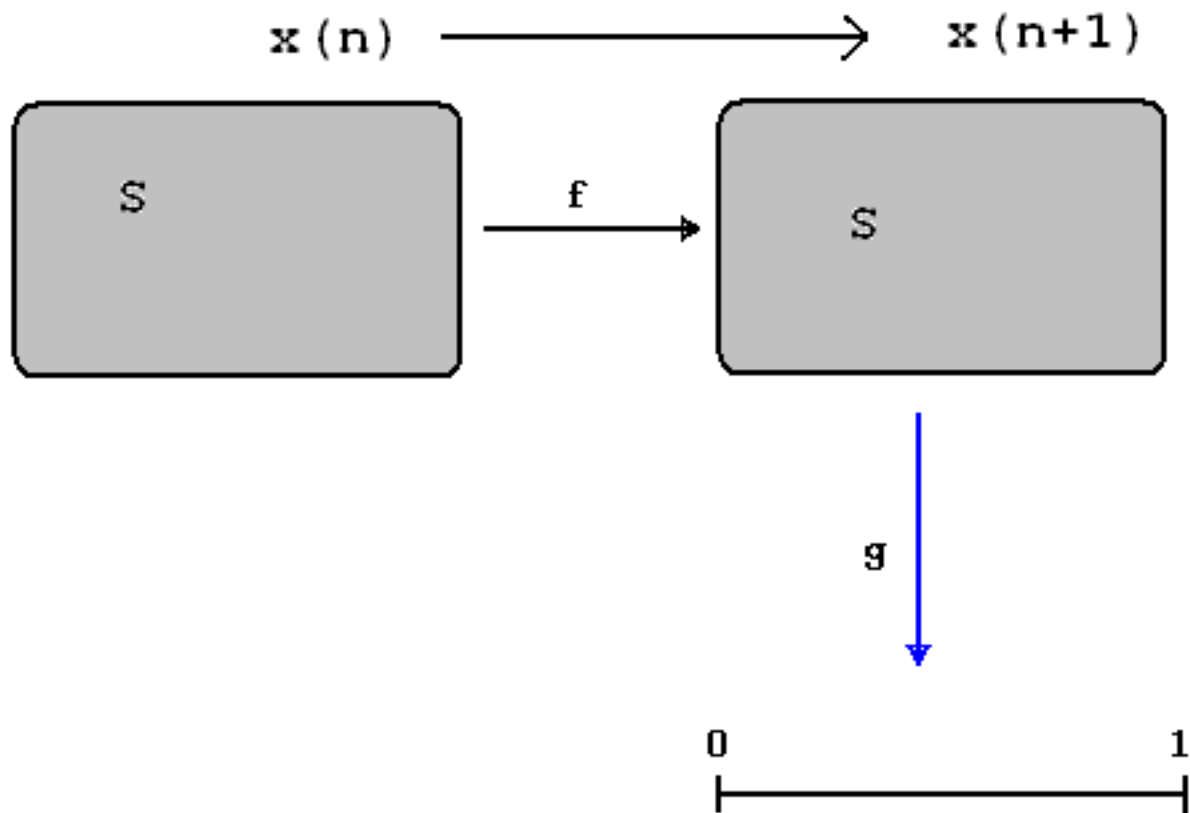
- Recursive rule $x_{n+1} = f(x_n)$ operating on fixed-length vectors x_n of integers, plus simple mapping $g : x_n \mapsto U_n$ so that U_1, U_2, \dots, U_N behaves like independent identically distributed Uniform[0,1] random variables

Classic example: $x_n = 0, \dots, 2^{31} - 1$ $U_n = x_n / 2^{31}$

- *Linear Congruential:* $x_{n+1} = a \cdot x_n + b \pmod{m}$

$$a = 7^5, b = 0, m = 2^{31} - 1$$

(Park & Miller, *Trans. ACM.* 1988)



Defining 'Simulation', cont'd

Ingredient #2: Expression of desired data structure:

Data as function of Building Block Uniform(0,1) r.v.'s

Examples: (a) Drawing from a list $1 \dots 23$ *with replacement*

`Uvec = runif(100)` gives 100-vector `Uvec`

which can be treated as indep. $\text{Unif}[0,1]$

`Xvec = trunc(23*Uvec) + 1`

`X = 1 + greatest integer \leq 23*U`

(b) How would you code 100 independent random selections from $1 \dots 230$ *with replacement* ?

(c) Selections of 100 from $1 \dots 230$ **without replacement** ?

More on Defining ‘Simulation’

Ingredient 2, cont’d: coding ‘data’ from indep. U_n

(d) **5-card poker hands:** 5 w.o. replacement from 1...52

```
Xvec = trunc(52*Uvec)+1           > Poker
Xnew = unique(Xvec)[1:5]         Clubs   Diam    Heart   Spad
Cards = 1+(Xvec-1) %% 13        "2.Cl"  "2.Di"  "2.He"  "2.Sp"
Hand = Poker[Xnew]              "3.Cl"  "3.Di"  "3.He"  "3.Sp"
Pairs = sum(table(Cards)==2)    "4.Cl"  "4.Di"  "4.He"  "4.Sp"
                                ...
```

In Hand of 5 cards, tabulate # pairs among card values 2...A

'Simulation', Ingredient 3

Question or Event specification or Variable to Average.

(e) **Geometric Probability:** what fraction of random points in the Unit Square fall in Inscribed Circle ?

Coding: Uvec, Vvec vectors of X and Y coordinates

Variable: DistSq = $(Uvec - .5)^2 + (Vvec - .5)^2$

Question: InCirc = $(DistSq < 1/4)$

Proportion in Circle = Area = $\pi/4 = .78540$

Average $Dist^2 = \int_0^1 \int_0^1 \{ (u - .5)^2 + (v - .5)^2 \} du dv = 1/6$

Data from Examples

Poker: Question is prob of 2 pairs, `xxyyz`

Combinatorial answer is: $\frac{1}{\binom{52}{5}} \binom{13}{2} \binom{4}{2} \binom{4}{2} 44 = 0.047539$

3 Runs, each with 10^5 simulated hands:

Run 1: 4793 of 1e5 had 2 pairs: estimated prob = .04793

Run 2: Tally of # pairs is :	0	1	2
	52669	42504	4827

Run 3: Tally of # pairs is :	0	1	2
	52880	42341	4779

Data from Geom. Prob Example

In successive runs of N randomly generated points in Unit Square:

Run#	N	Radius	InCirc	AvDistSq
1	1e5	.5	0.7871300	0.1662735
2	1e5	.5	0.7855700	0.1664821
3	1e6	.5	0.7849080	0.1668259
4	1e6	1/3	0.3491230	0.1667787
5	1e6	1/3	0.349002	0.166720

Worksheet Questions. #1. Find a single best estimate from these Data for the probability of a random point falling in the Inscribed Circle, of radius $1/2$ about $(1/2, 1/2)$?

#2. Can you account for the relative frequencies with which random points fall in the circle of radius $1/3$ about $(1/2, 1/2)$?

Conditional Prob's via Simulation

Conditional questions come up naturally:
condition determines denominator !

Example. Conditional prob. $X \in (.2, .6)$ given $Y \in (.3, .8)$:

(A) if (X, Y) random in the square

(B) if (X, Y) random in the circle $(X - .5)^2 + (Y - .5)^2 < 0.25$

(C) if (X, Y) random in the triangle $X < Y$

Simulations show the difference !

CondProb Demo

Further Worksheet Questions

- #3. What is the exact conditional probability of
(or relative area of region with)
 $X \in (.2, .6)$ given $Y \in (.3, .8)$ for a random point
 (X, Y) in the triangular region $0 \leq X < Y \leq 1$?
- #4. Since all of these simulations must be programmed:
how might one tell that there are errors in the program,
or that the random number generator is not behaving
properly ?

This is a probability related question: but we have not touched on the theoretical idea yet: that comes next.

Law of Large Numbers

If X_1, X_2, \dots, X_N are bounded random variables, independent and identically distributed, then

$$P(|(X_1 + \dots + X_N)/N - E(X_1)| > \epsilon) \rightarrow 0$$

as $N \rightarrow \infty$, for each $\epsilon > 0$.

Key example: $X_i = \{0, 1\}$ indicator that event E occurs in i 'th replicated dataset. Then $E(X_1) = P(E)$, heads probability.

So the LLN lets us make a prediction: if we think a simulation is erratic because of inadequate sample size, then it ought to settle down to stable results with larger N .

Large N Behavior of Estimate \hat{p}

Picture in [CumPoker Demo](#)

shows estimated fraction of points falling within circle of radius $1/9$ about $(1/2, 1/2)$ as number of points N in unit square grows.

To get quantitative idea of errors & variability in simulation averages for a particular N , we next appeal to the **Central Limit Theorem**.

(DeMoivre-Laplace) Central Limit Theorem

For fixed heads-probability p and number of independent identical-mechanism coin tosses, the random number of heads S_n among the first n tosses has **Binomial**(n, p) prob. distribution: $P(S_n = k) = \binom{n}{k} p^k (1-p)^{n-k}$, $0 \leq k \leq n$

For fixed heads-probability p and interval (a, b) , as $n \rightarrow \infty$:

$$P\left(a < \frac{S_n - np}{\sqrt{np(1-p)}} < b\right) \rightarrow \int_a^b e^{-x^2/2} \frac{dx}{\sqrt{2\pi}} \equiv \Phi(b) - \Phi(a)$$

Therefore with prob. $\Phi(b) - \Phi(a)$, $\hat{p}_n \equiv \frac{S_n}{n}$ lies between

$$\frac{1}{n} (np + a \sqrt{np(1-p)}) = p + a \sqrt{\frac{p(1-p)}{n}} \quad \text{and} \quad p + b \sqrt{\frac{p(1-p)}{n}}$$

Precision Bounds for Relative Frequencies

So if we simulate N replications E_1, \dots, E_n of event E

and use relative frequency $\hat{p}_N = \frac{1}{N} \sum_{i=1}^N I[E_i \text{ occurs}]$

to estimate $p = P(E)$, then

$$\frac{|\hat{p}_N - p|}{\sqrt{p(1-p)}} \quad \text{is bounded by} \quad \begin{cases} 1.96/\sqrt{N} & w.p. 0.95 \\ 2.576/\sqrt{N} & w.p. 0.99 \\ 3.291/\sqrt{N} & w.p. 0.999 \end{cases}$$

Even when true p is unknown, w.p. $\geq .999$, successive \hat{p}_N from separate simulation batches of size N cannot be farther apart than $\sqrt{4p(1-p)} \cdot 3.291/\sqrt{N} \leq 3.291/\sqrt{N}$

(This relates to Worksheet Question #4 above.)

Application of Precision Bounds

Recall data from 3 runs of 10^5 simulated Poker Hands:

Run 1 $\hat{p} = .04793$; **Run 2** $\hat{p} = .04827$; **Run 3** $\hat{p} = .04779$

With true $p \approx .048$, find 99% precision bounds

$$2.576\sqrt{(.048)(.952)/1e5} = 0.00174$$

(Multiply by $\sqrt{2}$ to bracket pairwise differences.)

Combine all three runs ($N=3e5$) by averaging, to get .04800

with .999 precision bound $3.291\sqrt{(.048)(.952)/3e5} = .00128$.

Exact 2-pair prob. = 0.047539, well within bounds.

$$(.04800 - .047539) / \text{sqrt}((.048) * (.952) / 3e5) = 1.181$$

is a perfectly unexceptional normal deviate.

Definitions: Density & Histogram

Probability Density: function $f \geq 0$, with $\int_{-\infty}^{\infty} f(x) dx = 1$
With random variable following density f

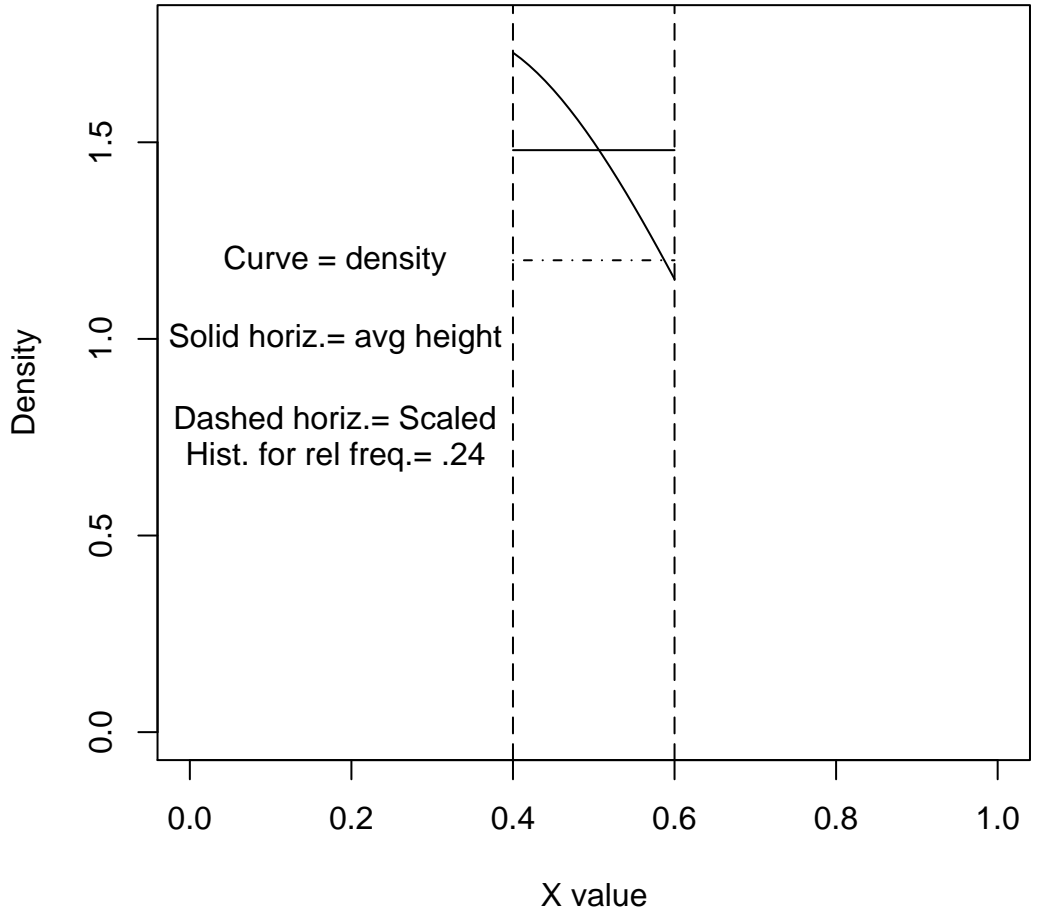
$$\text{Area under } f \text{ over } (a, b] = \int_a^b f(x) dx = P(a < X \leq b)$$

Scaled Rel. Freq. Histogram: based on counts n_1, n_2, \dots, n_L of **numbers of** variable values X_1, X_2, \dots, X_N resp. falling into (equal-length) intervals $(jh, (j+1)h]$.

Histogram: $g(x) = \frac{n_j}{Nh}$ for $jh < x \leq (j+1)h$

(Scaling makes total area under g equal to 1.)

Plot of Single-Cell Histogram Bar & Density Segment over the interval (0.4, 0.6]



Relationship: Density vs. Histogram

Suppose X_1, \dots, X_N data points, tallied for histogram,
with n_j values falling between $jh, (j+1)h$.

If f is true density for the X 's, then LLN says for large N :

$$\frac{n_j}{N} \approx P(jh < X_1 \leq (j+1)h) = \int_{jh}^{(j+1)h} f(x)dx$$

But the j 'th Scaled Histogram Bar is then

$$\frac{n_j}{Nh} \approx \frac{1}{h} \int_{jh}^{(j+1)h} f(x)dx = \text{Avg. Density Height in Cell}$$

which is close to $f(jh)$ when h is small !

Further Worksheet Problems

- #5. Suppose we do a simulation with $N = 2000$ iterations to evaluate a probability p which (an initial few simulations show) is in the neighborhood of 0.2. What is the 99% precision bound for the estimate (*i.e., the upper bound on $\hat{p} - p$ which holds with approximate probability 0.99*) ?
- #6. A certain type of density g is positive only on the interval $[0, 1]$ and has a constant value $g_j \leq 3$ on each of the intervals $(j/20, (j + 1)/20]$. Random variable values Y_1, \dots, Y_N are observed, with $N = 1000$. How accurate are the histogram bar heights as estimates of g_j , if you can tolerate a probability of error of 0.01 in your precision bounds ?

Permutational Hypothesis Tests

Student 1908 dataset: 20 values of shoe wear, in 2 groups.

Want to see if observed difference 0.41 of GroupB-GroupA averages is meaningfully large.

In the combined group of 20 shoe-wear values, there are $\binom{20}{10} = 184756$ ways to assign a subset of 10 as an artificial group B^* with A^* as its complement. Out of all ways, we want to know proportion giving

$$\text{abs}(B^* \text{ mean} - A^* \text{ mean}) > 0.41.$$

In this example could be found by enumeration, but generally only by sampling.

In `Demo PermShoe.txt`, we sample 20,000.

Efron's Nonparametric Bootstrap

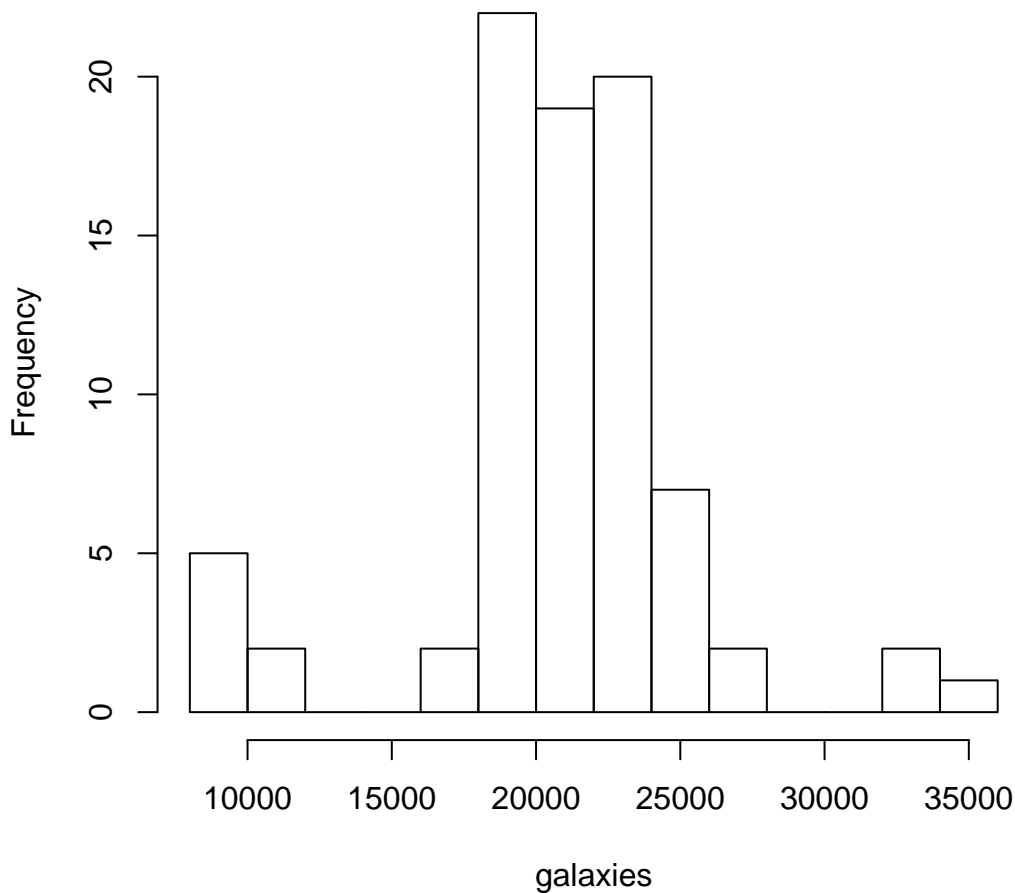
Illustrate another setting for randomly sampling from data.

Dataset of values of km/sec velocities of 82 galaxies.

Want to find IQR *Interquartile range* [61'st smallest minus 21'st smallest] and know its Standard Deviation (= square root of Variance) *for this moderate sample* **from this distribution.**

Histogram and results from 10,000 with-replacement samples of 82 values each from the same Galaxies dataset shown next:

Histogram of 82 Galaxies, IQR = 3385
Std. Dev. of 1e4 IQR's from Sample With Rep = 405.81



References

Google **Random Number Generation**

http://en.wikipedia.org/wiki/Random_number_generator

Diaconis, P. & Efron, B. (1983). Computer-intensive methods in statistics. *Scientific American* May, 116-130.

Lecture slides at: [http://www.math.umd.edu/ evs/MMIslides.pdf](http://www.math.umd.edu/evs/MMIslides.pdf) .

Visit the **R project** website <http://www.r-project.org/> for freely downloadable software !

Scripts for R code in demos at:

[http://www.math.umd.edu/ evs/MMIscriptR.txt](http://www.math.umd.edu/evs/MMIscriptR.txt)