- A. Monte Carlo (History and Overview)
- B. A Monte Carlo method to calculate π

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C. General Monte Carlo integration

Monte Carlo

 $\int_{a}^{b} f(x) dx$



 $\int_{a}^{b} f(x) dx$





We can approximate the area with thin bars that have all the same width, and so fill the area between a and b





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THE MONTE CARLO METHOD

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We shall present here the motivation and a general description of a method dealing with a class of problems in mathematical physics. The method is, essentially, a statistical approach to the study of differential equations, or more generally, of integro-differential equations that occur in various branches of the natural sciences.



Fig. 3. A subprogram written by Fermi for converting data in memory from hexadecimal to decimal form and printing the results.



Fig. 4. A subprogram written by Fermi for calculating phase shifts by finding a minimum chi-squared in a fit to the data.



Fig. 5. A portion of the printout of the program containing the subprograms described in Figs. 3 and 4. The program is written in machine language in hexadecimal

Monte Carlo method is tied to first electronic computers and the development of the atomic bomb.



Stan Ulam, Enrico Fermi, John von Neumann, Nicolas Metropolis, Edward Teller, Marshall Rosenbluth Augusta Harkanyi Teller, and Arianna Rosenbluth





Recipe

Do this many times:

Draw a random number x using the distribution f(x) between a and b
Save x

•Save x

Create a histogram of all the x values

THE ACCEPTANCE-REJECTION METHOD

Fig. 4. If two independent sets of random numbers are used, one of which (x^i) extends uniformly over the range of the distribution function f and the other (y^i) extends over the domain of f, then an acceptance-rejection technique based on whether or not $y^i \leq f(x^i)$ will generate a distribution for (x^i) whose density is $f(x^i) dx^i$.



Random numbers

Pseudorandom numbers

John von Neumann proposed using the following method as one of the first random number generators. Suppose we want to create eight-digit numbers. Begin with an eight-digit number X_0 , which we call the *seed*, and create the next integer in the sequence by removing the middle eight digits from X_0^2 .



Physical random number generators

Lava-lamps Geiger counter Devices that combine multiple events /dev/random



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Pseudocode for von Neuman random number generator

```
initialize vector x
seed = input(a number with 8 digits)
for i in 1,n:
    x0 = seed * seed
    seed = take middle 8 digits of x0
    x[i] = seed
```

Example

$\begin{array}{r} 81989672 \\ 81989672 \times 81989672 \\ 6722306314667584 \\ 30631466 \\ 30631466 \times 30631466 \end{array}$





Recipe

Do this many times:

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Create a histogram of all the x values



Recipe

Do this many times:

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Create a histogram of all the x values







Monte Carlo Rules

- Probability distribution functions (pdf's) the physical (or mathematical) system must be described by a set of pdf's.
- Random number generator a source of random numbers uniformly distributed on the unit interval must be available.
- Sampling rule a prescription for sampling from the specified pdf's, assuming the availability of random numbers on the unit interval, must be given.
- Scoring (or tallying) the outcomes must be accumulated into overall tallies or scores for the quantities of interest.

Monte Carlo Examples

Nuclear reactor design Quantum chromodynamics Radiation cancer therapy Traffic flow Stellar evolution **Econometrics Dow-Jones forecasting** Oil well exploration VLSI design

Phylogeny inference Population genetics inference

?

We know that the area of a circle is

 πr^2

r r

Looking only at the upper right corner we can see a green square with side r and we can calculate the area of the square as

$$A_s = r^2$$

The quarter circle has the area

$$A_c = \frac{\pi}{4}r^2$$

So we can calculate the ratio of the two areas as

$$\frac{A_c}{A_s} = \frac{r^2}{\frac{\pi}{4}r^2} = \frac{\pi}{4}$$



$$\frac{A_c}{A_s} = \frac{r^2}{\frac{\pi}{4}r^2} = \frac{\pi}{4}$$

The goal is now to estimate the ratio of the areas. We can devise an algorithm that draws random coordinates from the square and marks whether the coordinate fell into the circle or not.We can calculate the distance from the circle center using Pythagoras:

$$d = \sqrt{(x^2 + y^2)}$$

If d is smaller than r than we know the coordinate is in the circle otherwise only in the square. We can now create an algorithm for our program.



// Algorithm in pseudo code
// Do many times:
// draw x, y coordinate
// calculate d from center
// check whether d < r:
// True: add 1 to circle
// False: do nothing
// add 1 to square</pre>

// print pi: ratio circle/square * 4

Our Pi estimates



 $\hat{\pi}$

History of π

Archimedes (300 BC) using 96-side polygons



223/71 < π < 22/7 3.140845070422535 3.1428571428571428

Ptolemy	(c. 150 AD)	3.1416
Zu Chongzhi	(430-501 AD)	³⁵⁵ / ₁₁₃
al-Khwarizmi	(c. 800)	3.1416
al-Kashi	(c. 1430)	14 places
Viète	(1540-1603)	9 places
Roomen	(1561-1615)	17 places
Van Ceulen	(c. 1600)	35 places

James Gregory $\frac{\pi}{4} = 1 - \frac{1}{3} + \frac{1}{5} - \frac{1}{7} + \dots$

History of π

1699: Sharp used Gregory's result to get 71 correct digits
1701: Machin used an improvement to get 100 digits and the following used his methods:
1719: de Lagny found 112 correct digits
1789: Vega got 126 places and in 1794 got 136
1841: Rutherford calculated 152 digits and in 1853 got 440
1873: Shanks calculated 707 places of which 527 were correct

Very soon after Shanks' calculation a curious statistical freak was noticed by De Morgan, who found that in the last of 707 digits there was a suspicious shortage of 7's. He mentions this in his Budget of Paradoxes of 1872 and a curiosity it remained until 1945 when Ferguson discovered that Shanks had made an error in the

528th place, after which all his digits were wrong. In 1949 a computer was used to calculate π to 2000 places. In this and all subsequent computer expansions the number of 7's does not differ significantly from its expectation, and indeed the sequence of digits has so far passed all statistical tests for randomness.

Buffon's needle experiment. If we have a uniform grid of parallel lines, unit distance apart and if we drop a needle of length k < 1 on the grid, the probability that the needle falls across a line is $2k/\pi$. Various people have tried to calculate π by throwing needles. The most remarkable result was that of Lazzerini (1901), who made 34080 tosses and got $\pi = 355/113 = 3.1415929$

which, incidentally, is the value found by Zu Chongzhi. This outcome is suspiciously good, and the game is given away by the strange number 34080 of tosses. Kendall and Moran comment that a good value can be obtained by stopping the experiment at an optimal moment. If you set in advance how many throws there are to be then this is a very inaccurate way of computing π . Kendall and Moran comment that you would do better to cut out a large circle of wood and use a tape measure to find its circumference and diameter.

π

In the State of Indiana in 1897 the House of Representatives unanimously passed a Bill introducing a new mathematical truth:

Be it enacted by the General Assembly of the State of Indiana: It has been found that a circular area is to the square on a line equal to the quadrant of the circumference, as the area of an equilateral rectangle is to the square of one side. (Section I, House Bill No. 246, 1897)

The Senate of Indiana showed a little more sense and postponed indefinitely the adoption of the Act!

3 14150265358070323846264338327050288410716030037510582007404450230781640628620800

Using Monte Carlo to approximate an integral

• Suppose we want to evaluate $\int_{a}^{b} f(x) dx$

- If $f(x) \ge 0$ for $a \le x \le b$ then we know that this integral represents the area under the curve y = f(x) and above the x-axis.
- Standard deterministic numerical integration rules approximate this integral by evaluating the integrand f(x) at a set number of points and multiplying by appropriate weights.

- For example, the midpoint rule is

$$\int_{a}^{b} f(x) \, dx \approx f\left(\frac{a+b}{2}\right)(b-a)$$

- Simpson's rule is

$$\int_{a}^{b} f(x) \, dx \approx \frac{b-a}{6} \Big[f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \Big]$$

THE ACCEPTANCE-REJECTION METHOD

Fig. 4. If two independent sets of random numbers are used, one of which (x^i) extends uniformly over the range of the distribution function f and the other (y^i) extends over the domain of f, then an acceptance-rejection technique based on whether or not $y^i \leq f(x^i)$ will generate a distribution for (x^i) whose density is $f(x^i) dx^i$.



Acceptance-Rejection method

Area

 $\Lambda_{i} = \Lambda I (h a)$



Area under curve
$$A_b = M(b-a)$$

 $A_b = M(b-a)$
 $\#inside$
 $\#inside$
 $\#bounding box$

Downding how

We need to be able to calculate f(x) for any possible x with the range a and b.

We draw 2 random values, one for x and one for y. then evaluate f(x), if y < f(x) then we count this as #inside. The #bounding box is the total number of draws.

Problem with the Acceptance-Rejection method

- How to choose the bounding box? It need to be big enough to contain the whole function. But if it is too big then we draw often random numbers above the function. If the bounding box is much larger than the area under the curve then we need many draws (or steps) to get a good accuracy.
- We need to draw two random numbers and 'discard' the draws that are above the function.

Usual Monte Carlo Integration

$$y = f(x)$$

$$A = \int_{a}^{b} f(x)dx$$

$$B = (b - a)f(c)$$

$$A = \int_{a}^{b} f(x)dx$$

$$B = (b - a)f(c)$$
There must be an $f(c)$ that satisfies
$$A = B.$$
Find $f(c)$ and we are done!

where x_i are drawn uniformly between a and b

Monte Carlo Integration Algorithm

- Draw many x_i between the boundaries a and b
- Calculate the average \overline{x} of the collected x_i
- Calculate the area as (b-a) \overline{x}

Evaluation of Monte Carlo error

We will discuss how to calculate error at the end of the semester when we again talk about Monte Carlo.