

## What is Monte Carlo Simulation ?

- **Monte Carlo methods** are a widely used class of **computational algorithms** for simulating the behavior of various physical and mathematical systems, and for other computations.
- **Monte Carlo algorithm** is often used to find solutions to mathematical **numerical Monte Carlo method** problems (which may have many variables) that cannot easily be solved, (e.g. integral calculus, or other numerical methods)

## Monte Carlo Simulation

- A scheme employing random numbers which is used to solve certain stochastic or deterministic problems where the passage of time plays no substantive role.
- Common problem is estimation of  $\int_{\Omega} f(\mathbf{x}) d\mathbf{x}$ , where  $f$  is a function,  $\mathbf{x}$  is a vector and  $\Omega$  is domain of integration.
- Special case: Estimate  $\int_a^b f(x) dx$  for scalar  $x$  and limits of integration  $a, b$

## Monte Carlo Simulation

Let  $X$  be a uniform random variable on the interval  $[a, b]$  with density

$$p(x) = \frac{1}{b-a}, \quad a \leq x \leq b$$

and let  $x_1, \dots, x_n$  be a random sample from  $X$ .

Then

$$\begin{aligned} \int_a^b f(x) dx &= \int_a^b \frac{f(x)}{p(x)} p(x) dx \\ &= (b-a) \int_a^b f(x) p(x) dx \\ &= (b-a) E[f(X)] \\ &\approx \frac{b-a}{n} \sum_{i=1}^n f(x_i). \end{aligned}$$

3

## Monte Carlo Simulation

Example: Estimate  $\int_0^b \sin(x) dx$ .

We approximate this by

$$\frac{b}{n} \sum_{i=1}^n \sin(x_i),$$

where  $x_1, \dots, x_n$  are a sample from a uniform  $[0, b]$  random variable.

4

## Monte Carlo Simulation

Example: Estimate  $\int_0^b \sin(x)dx$ .

	$n = 10$	$n = 100$	$n = 1000$	$n = 2000$
$b = \pi$ (answer = 2)	1.753	2.032	1.994	1.999
$b = 2\pi$ (answer = 0)	-0.898	-0.013	0.137	0.079

There is considerable variability in the quality of solution; accuracy of numerical integration sensitive to integrand and domain of integration

5

## Case Study

### Cake's shop problem

An owner of a bakery shop would like to determine how many 10-inch birthday cakes he should produce each day in order to maximize his profit. His present method of determining the quantity to bake is based on his best guess.

## Cake's shop problem

- The production costs are \$2.00 per cake.
- And the profit for each cake is \$2.5.
- However, If over estimates the daily demand, some cakes will be left over at the end of the day. The policy is to sell all leftover cakes to a local store that specializes in day-old items. He is currently receiving \$1.50 per cake for the surplus cakes, thus incurring a loss of \$0.50 per cake.

## Cake's shop problem

- Case 1: The production quantity is less than or equal to demand

$$\text{If } x \leq d, \quad z = 2.5x$$

- Case 1: The production quantity is greater than the demand

$$\text{If } x > d \quad z = 2.5d + (x-d) (-0.5)$$

$$Z = 3.00 d - 0.5 x$$

## Cake's shop problem

- Generalizeing:

p = selling price for each cake

c = cost of each unit

s = day-old price

- If  $x \leq d$ ,  $z = (p-c) x$

- If  $x > d$   $z = (p-c) d + (x-d) (s-c)$

$$Z = (p-s) d - (s-c) x$$

## Historical day demand for the birthday cakes

Daily deman	Frequency	Relative freq
0	1	0.05
1	2	0.1
2	1	0.05
3	2	0.1
4	3	0.15
5	6	0.3
6	3	0.15
7	1	0.05
8	1	0.05
Total	20	1

$$relative\_frequency = \frac{frequency\_of\_observation}{total\_number\_of\_observations}$$

## Hand Simulation

- Take a sheet of paper and cut it into twenty equal pieces.
- Follow the historical daily demand frequency in the table,
- write the number zero on one piece.
- On two of the remaining pieces write the number one, which stands for the demand of one unit.
- Check the numbers you have written carefully, because this “deck” of twenty

## Hand Simulation

- The first step is the selection of the production quantity, Assume ( $x=3$ ).
- use the deck of twenty slips of paper to generate a demand by selecting one slip of paper at random.
- Suppose the first slip drawn has a 5 written on it.
- We shall then use a demand of 5 cakes for the first simulated day of bakery shop operation.
- i.e. underproduction of 2 cakes.

## Hand Simulation

- Since  $x < d$ , we can compute our first day's profit using the expression
- $2.5x = 2.5(3) = \$7.5$ . i.e. Total profit of \$7.5.
- Generate the demand for second day (reshuffle and draw a piece) suppose  $d=1$
- Since  $x > d$  use the second case
- $z = 3(1) - 0.5(3) = \$1.5$
- So the total profit is  $7.5 + 1.5 = 9$

## 10-day simulation results for production quantity $x=3$

Day	Generated demand	Daily profit	Total profit
1	5	7.5	7.5
2	1	1.5	9
3	6	7.5	16.5
4	3	7.5	24
5	4	7.5	31.5
6	4	7.5	39
7	3	7.5	46.5
8	0	-1.5	45
9	5	7.5	52.5
10	6	7.5	60

## Hand Simulation

- Now we perform the same ten day simulation for another quantity production  $x = 1, 2, 3 \dots 8$
- Compare the total profit for each one
- Pick the best profit to be the suggested production quantity
- Of course if we run the simulation for more days we get more accurate estimate.

## 10-days Simulation Results for various production quantities

Production Size	Ten Day Simulated profit \$
1	25
2	44
3	60
4	79
5	90
6	93
7	91
8	89
9	5
10	6

From the table it is clear that the best production quantity that maximizes the profit is at  $x=6$ . The results are based on only 10-day simulation.



## The role of random numbers in simulation

- Suppose we select random numbers in sets of two digits.
- This will provide us with 100 two-digit random numbers from 00 to 99 with each two-digit random number having a  $1/100$  chance of being selected
- 0 unites, the relative frequency of 0 is 5% Thus we want 5% of the 100 possible two-digit random numbers to correspond to a demand of 0 units.
- While choosing any five numbers of the 100 numbers will do we may assign a demand of 0 to the first 5 numbers i.e. 00, 01, 02, 03, and 04

## Random number Intervals and the daily demand

Daily Dema	relative Frequen	Interval of Random num
0	0.15	00 to 04
1	0.1	05 to 14
2	0.05	15 to 19
3	0.1	20 to 29
4	0.15	30 to 44
5	0.3	45 to 74
6	0.15	75 to 89
7	0.05	90 to 94
8	0.05	95 to 99

## Results of simulating ten daily demands

Random number	Simulated daily demand	simulated daily demand
63		5
27		3
15		2
99		8
86		6
71		5
74		5
45		5
11		1
2		0

## The role of random numbers

- For any simulation problem in which a relative frequency distribution of a variable can be developed,
- It is easy to apply the above random number based procedure to simulate values of the variable.
- First, develop a table of intervals by associating an interval of random numbers with each possible value of the variable
- Then as each random number is selected, you can simply check the corresponding interval and find the associated value of the variable.

## The role of random numbers

- Obviously, for long and complex simulations that require numerous calculations, a high speed computer simulation process is desirable.
- In computer simulation pseudo- random numbers are used in exactly the same way as the random numbers selected from random number tables above.
- It would be very risky to make a decision based on the results of such a short period of simulation.

## The role of random numbers

- When we think of performing the simulation calculations for a simulated period as long as 500 days, the problems of carrying out the simulation for even a case as small as the bakery shop problem are significant.
- For example let us consider the 500 days. The mathematical model does not change but the work we have to go through to evaluate the results does change but expands . Now we can create a table similar to the ten-day table to evaluate each order size for 500 days of operation.