

# Monte Carlo method

モデル化とシミュレーション特論  
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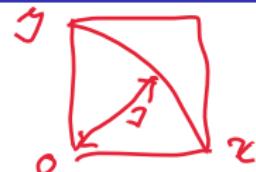
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# Monte Carlo method

- General framework for simulations using random numbers
- Numerical integrals
- Modeling stochastic processes
- Approximated solutions for difficult combinatorial optimizations
- sample code in the following URL

[https://github.com/modeling-and-simulation-mc-saga/  
MonteCarlo](https://github.com/modeling-and-simulation-mc-saga/MonteCarlo)

# Estimation of $\pi$



- Generate 2-dimensional random numbers in a square with unit length sides.

$$(x, y) \in [0, 1) \times [0, 1) \quad (2.1)$$

- Count events entering an arc with unit length radius.

$$0 \leq (x^2 + y^2)^{1/2} < 1 \quad (2.2)$$

- The probability of events entering the arc is  $\pi/4$

- $P_N(m)$ : Probability of  $m$  out of  $N$  points falling into the arc.

$$\underline{P_N(m)} = \binom{N}{m} p^m (1-p)^{N-m}, \quad p = \frac{\pi}{4} \quad (2.3)$$

- Probability generating function for  $P_N(m)$

$$\underline{G(z) = \sum_{m=0}^N P_N(m) z^m = (pz + 1 - p)^N} \quad (2.4)$$

$$G'(z) = \sum_{m=0}^N m P_N(m) z^{m-1} = N p (pz + 1 - p)^{N-1} \quad (2.5)$$

$$\langle m \rangle = G'(1) = \underbrace{N p}_{\text{---}} \quad (2.6)$$

$$\begin{aligned} G''(z) &= \sum_{m=0}^N m(m-1) P_N(m) z^{m-2} \\ &= N(N-1)p^2 (pz + 1 - p)^{N-2} \end{aligned} \quad (2.7)$$

$$G''(1) = \langle m^2 \rangle - \langle m \rangle = N(N-1)p^2 \quad (2.8)$$

$$\sigma^2 = \langle m^2 \rangle - \langle m \rangle^2 = N p (1 - p) \quad (2.9)$$

# Simulation

$$\frac{\langle m \rangle_{\text{exp}}}{N} \sim \frac{\pi}{4} \quad (2.10)$$

→  $\frac{\sigma}{\langle m \rangle} = \frac{1}{N^{1/2}} \left( \frac{1-p}{p} \right)^{1/2}$  (2.11)

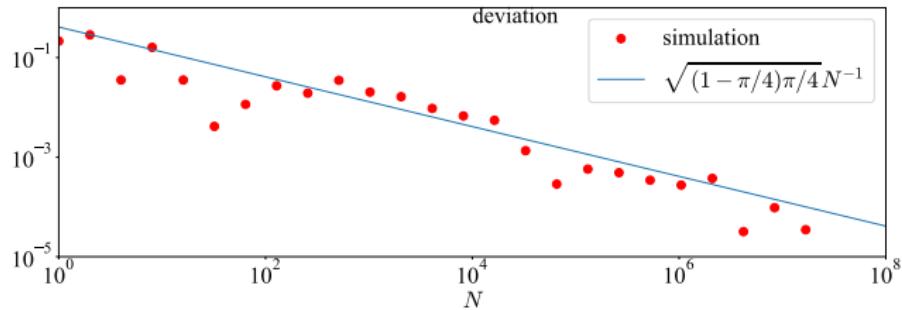
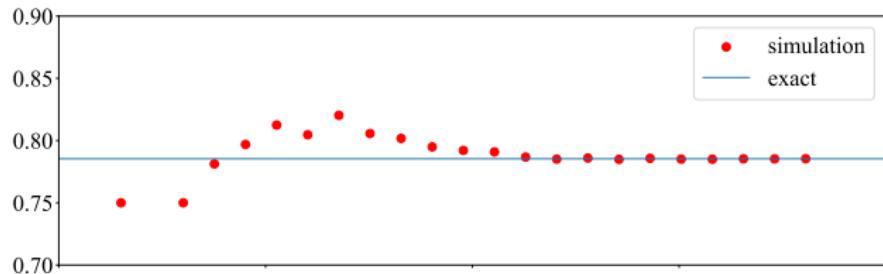
Deviation

$$\left| \frac{\langle m \rangle_{\text{exp}}}{N} - \frac{\pi}{4} \right| \quad (2.12)$$

will reduce with  $N^{-1/2}$

```
1 public double addOne() {  
2     all++;  
3     double x = myRandom.nextDouble();  
4     double y = myRandom.nextDouble();  
5     double r = Math.sqrt(x * x + y * y);  
6     if (r <= 1.) {  
7         in++;  
8     }  
9     return (double) in / all;  
10 }
```

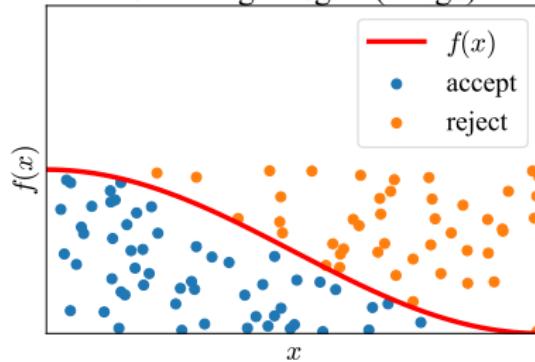
estimatingPi/Pi.java

Estimating  $\pi$ 

# Estimating Integrals

- Function  $f(x) \geq 0$  defined in  $x \in [a, b]$
- $m > \max f(x)$
- Generate  $n$  2-dimensional random numbers in  $[a, b] \times [0, m)$
- $k$  random numbers fallen inside the area under  $f(x)$
- The integral is  $(k/n) \times m \times (b - a)$

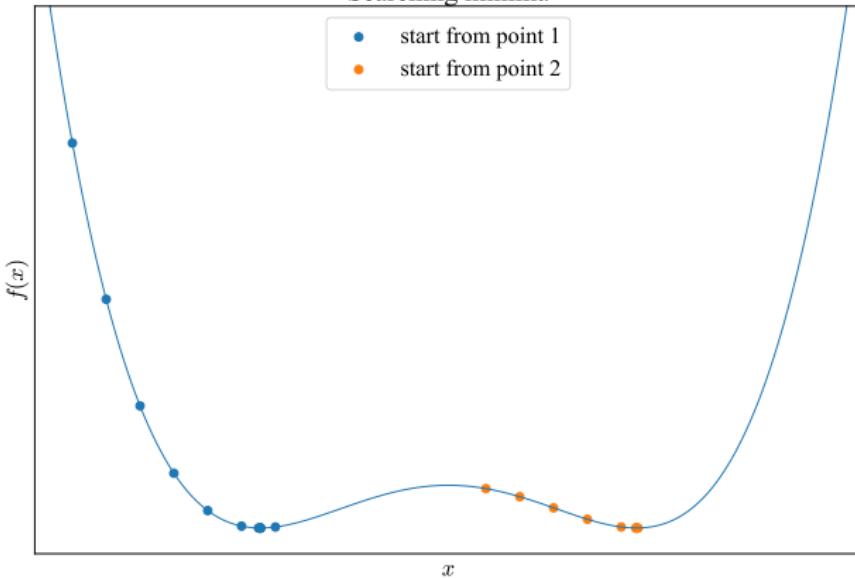
Estimating Integral (image)



# Find minima of $f(x)$

- There is only one minimum of  $f(x)$ .
  - Starting from an arbitrary  $x$ .
  - Change  $x$  slightly for reducing  $f(x)$
- There are some minima of  $f(x)$ .
  - Trials with randomly selected starting points.
- For cases of huge number of minima of  $f(x)$ ?

## Searching minima



# Random Spins

- $n$  variables  $s_i = \pm 1$
- Interactions  $J_{ij}$  ( $J_{ij} = J_{ji}, J_{ii} = 0$ )
  - Positive and negative random values
- Minimize

$$E = - \sum_{ij} J_{ij} s_i s_j \tag{5.1}$$

- Split  $s_i$  into two groups
- Example of bi-partitioning problems

# Monte Carlo method for spin systems

- Select randomly one  $s_i$  of spins, and try to flip it.

$$s_i \rightarrow s_i + \Delta x_i \quad (5.2)$$

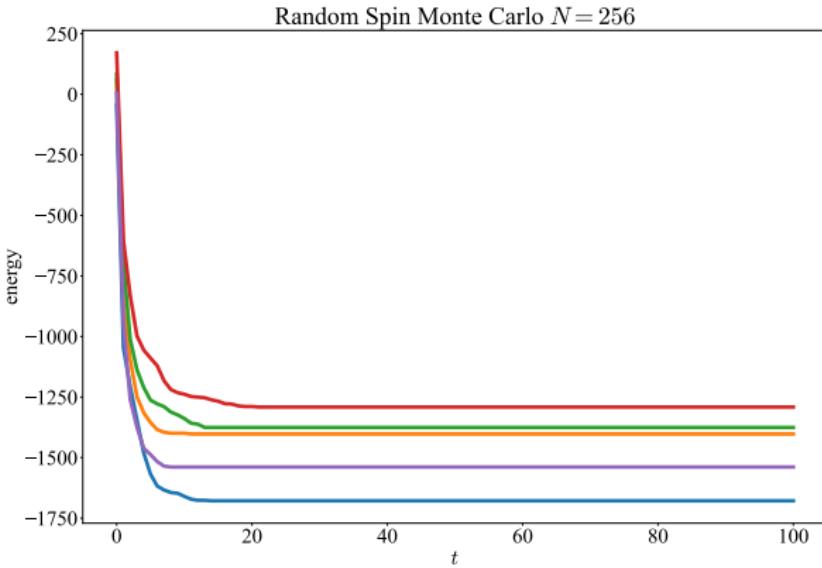
- Evaluate energy change by flipping the spin  $s_i$

$$\Delta E = -2 \sum_j J_{ij} s_j \Delta x_i \quad (5.3)$$

- If  $\Delta E < 0$ , then employ the new value for  $s_i$
- 1 Monte Carlo step :  $n$  trials for changing spins

```
1 public double oneStep() {
2     int k = random.nextInt(n); //Random selection of a spin
3     int ds = -2 * s[k]; //spin flip
4     //energy
5     double de = 0.;
6     for (int j = 0; j < n; j++) {
7         de += -2 * J[k][j] * s[j] * ds;
8     }
9     //the spin flips if energy decreases
10    if (de < 0) {
11        energy += de;
12        s[k] += ds;
13    }
14    return de;
15 }
```

simpleMonteCarlo/SpinSystem.java



- 5 different initial states lead to different states.
- This shows the existence of many energy minima
- How to find the minimum?

# DLA: Diffusion Limited Aggregation

- At the beginning, there is a seed of a cluster at the center.
- particles are diffused and adhered to the cluster if the particle contacts to the cluster.

# Simplest Simulation

- 2-dimensional lattice
- At the beginning, there is a seed of a cluster at the center.
- A particle enter the system, and do simple random walk.
  - one of four direction is randomly selected
  - if the position is next to the cluster, the particle is adhered.
  - A particle is adhered to the cluster, another particle enters the system.

<https://github.com/modeling-and-simulation-mc-saga/DLA>

```
1 public Point inject() {
2     Point p = new Point(0, 0);
3     setNewPosition(p); //Place a particle on the boundary
4
5     while (true) { //Repeat until absorption
6         if (isAdjacent(p)) { //The particle is absorbed
7             cells[p.x][p.y] = 1;
8             return p;
9         }
10        //Random displacement
11        int k = random.nextInt(4);
12        switch (k) {
13            case 0:
14                p.translate(1, 0);
15                break;
16            case 1:
17                p.translate(0, 1);
18                break;
19            case 2:
20                p.translate(-1, 0);
21                break;
22            default:
23                p.translate(0, -1);
24                break;
25        }
26        adjustPosition(p);
27    }
28}
```

30,000 particles

