

# Monte Carlo method

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

- General name for simulations using random numbers
- Numerical integrals
- Modeling stochastic processes
- Approximated solutions for difficult combinatorial optimizations

Estimation of  $\pi$ 

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

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

- Count when entering an arc with unit radius.

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

- The probability is  $\pi/4$

- Probability  $P_N(m)$  for  $m$  of  $N$  points entering the arc.

$$P_N(m) = \binom{N}{m} p^m (1-p)^{N-m}, \quad p = \frac{\pi}{4} \quad (3)$$

- Probability generating function

$$G(z) = \sum_{m=0}^N P_N(m) z^m = (pz + 1 - p)^N \quad (4)$$

$$G'(z) = \sum_{m=0}^N m P_N(m) z^{m-1} = Np (pz + 1 - p)^{N-1} \quad (5)$$

$$\langle m \rangle = G'(1) = Np \quad (6)$$

$$\begin{aligned} G''(z) &= \sum_{m=0}^N m(m-1) P_N(m) z^{m-2} \\ &= N(N-1)p^2 (pz + 1 - p)^{N-2} \end{aligned} \quad (7)$$

$$G''(1) = \langle m^2 \rangle - \langle m \rangle = N(N-1)p^2 \quad (8)$$

$$\sigma^2 = \langle m^2 \rangle - \langle m \rangle^2 = Np(1-p) \quad (9)$$

## Simulation

$$\frac{\langle m \rangle_{\text{exp}}}{N} \sim \frac{\pi}{4} \quad (10)$$

$$\frac{\sigma}{\langle m \rangle} = \frac{1}{N^{1/2}} \left( \frac{1-p}{p} \right)^{1/2} \quad (11)$$

Deviation

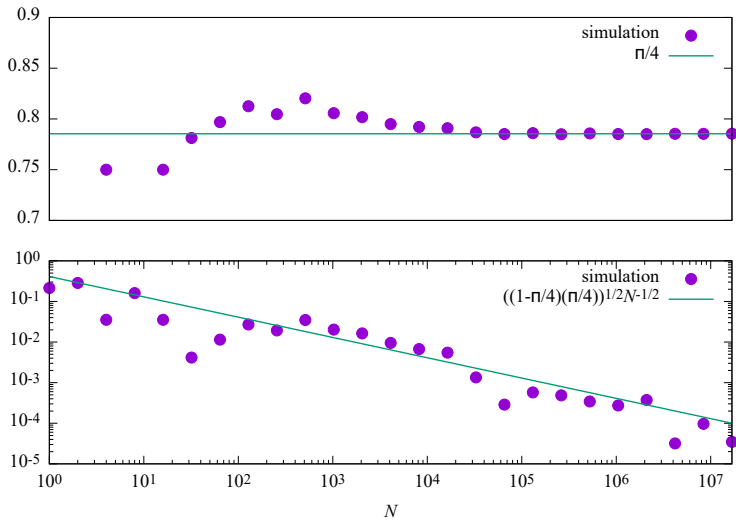
$$\left| \frac{\langle m \rangle_{\text{exp}}}{N} - \frac{\pi}{4} \right| \quad (12)$$

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 }
```

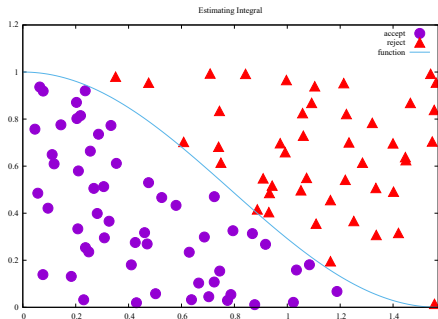
estimatingPi/Pi.java



Estimating  $p$ 

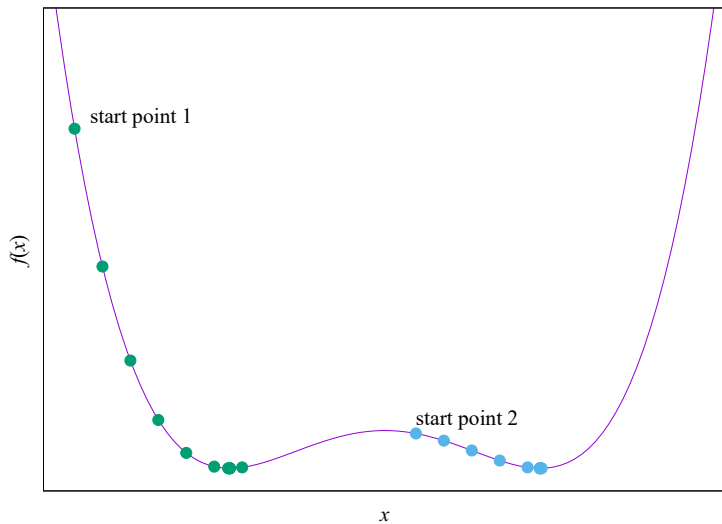
# Estimating Integrals

- Function  $f(x)$  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 fall inside the area under  $f(x)$
- The integral is  $(k/n) \times m \times (b - a)$



Find minima of  $f(x)$ 

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



# Random Spins

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

$$E = - \sum_{ij} J_{ij} s_i s_j \quad (13)$$

- Split  $s_i$  into two groups
- Example of two-part problem

# Monte Carlo method for spin systems

- Randomly select one spin  $s_i$ , and try to flip it.

$$s_i \rightarrow s_i + \Delta x_i \quad (14)$$

- Evaluate energy change by flipping the spin

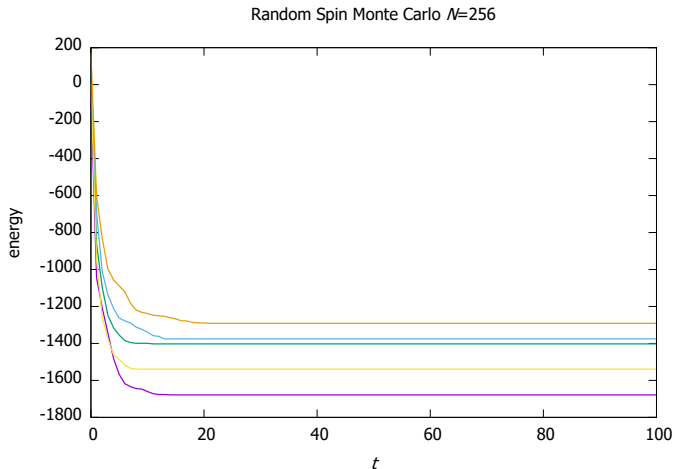
$$\Delta E = -2 \sum_j J_{ij} s_j \Delta x_i \quad (15)$$

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); //ランダムにスピンを生成
3     int ds = -2 * s[k]; //スピンを反転する大きさ
4     //エネルギー変化を計算
5     double de = 0.;
6     for (int j = 0; j < n; j++) {
7         de += -2 * J[k][j] * s[j] * ds;
8     }
9     //エネルギーが下がる場合に、スピンを本当に反転する
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



# 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); //境界に粒子を設定
4
5      while (true) { //吸着するまで繰り返す
6          if (isAdjacent(p)) { //クラスタに吸着
7              cells[p.x][p.y] = 1; return p;
8          }
9          //ランダムに移動
10         int k = random.nextInt(4);
11         switch (k) {
12             case 0:
13                 p.translate(1, 0); break;
14             case 1:
15                 p.translate(0, 1); break;
16             case 2:
17                 p.translate(-1, 0); break;
18             default:
19                 p.translate(0, -1); break;
20         }
21         adjustPosition(p);
22     }
23 }

```

30,000 particles

