Advanced topics in Markov-chain Monte Carlo

Lecture 1:

Transition matrices - from the balance conditions to mixing Part 1/2: Introduction

Werner Krauth

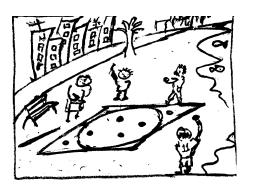
ICFP -Master Course Ecole Normale Supérieure, Paris, France

18 January 2023

References

 W. Krauth "Statistical Mechanics: Algorithms and Computations" (Oxford University Press, 2006; second edition: "soon")

Monte Carlo



Direct sampling (algorithm)

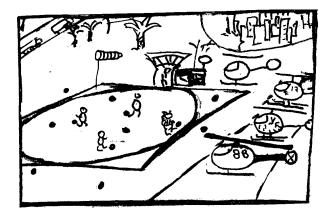
```
\label{eq:procedure} \begin{split} & \text{procedure} \text{ direct-pi} \\ & \textit{N}_{\text{hits}} \leftarrow 0 \text{ (initialize)} \\ & \text{for } i = 1, \dots, N \text{ do} \\ & \left\{ \begin{array}{l} x \leftarrow \text{ran}[-1, 1] \\ y \leftarrow \text{ran}[-1, 1] \\ \text{ if } (x^2 + y^2 < 1) \text{ $N_{\text{hits}}} \leftarrow N_{\text{hits}} + 1 \end{array} \right. \\ & \text{output $N_{\text{hits}}$} \end{split}
```

Direct sampling (results)

Five trials with N = 4000

run	$N_{\rm hits}$	estimation	
1	3156	3.156	
2	3129	3.129	
3	3154	3.154	
4	3134	3.134	
5	3148	3.148	

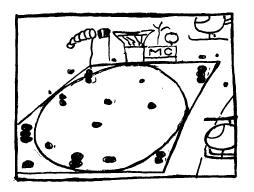
Markov-chain sampling (1/3)



Markov-chain sampling (2/3)

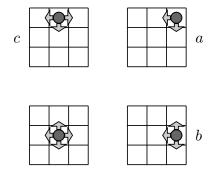
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procedure markov-pi
hits \leftarrow 0: x \leftarrow 1: v \leftarrow -1
for i = 1, ..., N do
     \begin{cases} \delta x \leftarrow \operatorname{ran}[-\delta, \delta] \\ \delta y \leftarrow \operatorname{ran}[-\delta, \delta] \\ \text{if (}|x + \delta x| < 1 \text{ and } |y + \delta y| < 1 \text{) then} \\ \begin{cases} x \leftarrow x + \delta x \\ y \leftarrow y + \delta y \\ \text{if (}x^2 + y^2 < 1 \text{) } N_{\text{hits}} \leftarrow N_{\text{hits}} + 1 \end{cases}
output N<sub>hits</sub>
```

Markov-chain sampling (3/3)



Metropolis et al. (1953).

3×3 pebble game



discretized version of heliport game

Global balance—detailed balance

$$\underbrace{p(a \rightarrow a)}_{\text{probability to go}} + p(a \rightarrow b) + p(a \rightarrow c) = 1$$

$$\underbrace{p(a \rightarrow a)}_{\text{from } a \text{ to } a}$$

global-balance condition

$$\underbrace{\pi(a)}_{\text{probability to be}} = \pi(a)p(a \to a) + \pi(b)p(b \to a) + \pi(c)p(c \to a)$$

$$\pi(a)p(a \rightarrow c) + \pi(a)p(a \rightarrow b) = \pi(b)p(b \rightarrow a) + \pi(c)p(c \rightarrow a)$$

detailed-balance condition

$$\pi(a)p(a \rightarrow b) = \pi(b)p(b \rightarrow a)$$
 etc

A priori probabilities, acceptance, implementation

			second		
7	8	9	Д		
4	5	6	third → → first		
1	2	3	. ↓		
			fourth		

Nbr(., <i>k</i>)					
1	2	3	4		
2	4	0	0		
3	5	1	0		
0	6	2	0		
5	7	0	1		
6	8	4	2		
0	9	5	3		
8	0	0	4		
9	0	7	5		
0	0	8	6		
	1 2 3 0 5 6 0 8 9	1 2 2 4 3 5 0 6 5 7 6 8 0 9 8 0 9 0	1 2 3 2 4 0 3 5 1 0 6 2 5 7 0 6 8 4 0 9 5 8 0 0 9 0 7		

Transition matrix

• *P*: matrix of conditional transition probabilities from *i* to *j*:

$$P = \begin{bmatrix} \frac{1}{2} & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot & \cdot & \cdot & \cdot & \cdot \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot & \cdot & \cdot & \cdot \\ \cdot & \frac{1}{4} & \frac{1}{2} & \cdot & \cdot & \frac{1}{4} & \cdot & \cdot & \cdot \\ \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot & \cdot \\ \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & 0 & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot \\ \cdot & \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & \frac{1}{4} & \cdot & \cdot & \frac{1}{4} \\ \cdot & \cdot & \cdot & \frac{1}{4} & \cdot & \cdot & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \cdot & \cdot & \cdot & \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{bmatrix}$$

Algorithmic probabilities II

initial probability vector

$$\pi^{\{t=0\}} = \{0, \dots, 0, 1\}.$$

probability at iteration i + 1 from iteration i

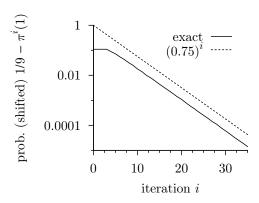
$$\pi_i^{\{t+1\}} = \sum_{j=1}^9 \pi_j^{\{t\}} P_{j \to i}$$

eigenvectors and eigenvalues

$$\{\pi_1^{\{t\}}, \dots, \pi_9^{\{t\}}\} = \underbrace{\{\frac{1}{9}, \dots, \frac{1}{9}\}}_{ \text{first (left) eigenvector eigenvalue } \lambda_1 = 1 } + \alpha_2 (0.75)^t \underbrace{\{-0.21, \dots, 0.21\}}_{ \text{second (left) eigenvector eigenvalue } \lambda_2 = 0.75 } + \dots$$

Transition matrices

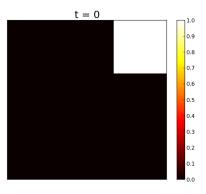
convergence of pebble game:



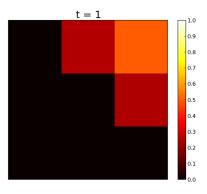
exponential convergence = scale

$$(0.75)^t = \exp[t \cdot \log \ 0.75] = \exp\left[-\frac{t}{3.476}\right].$$

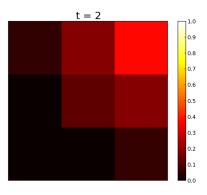
• Distribution $\pi^{t=0}$ (starting from upper right)



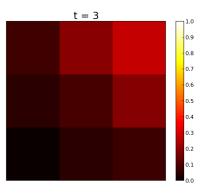
• Distribution $\pi^{t=1}$ (starting from upper right)



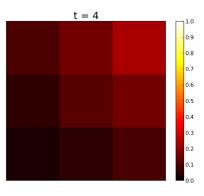
• Distribution $\pi^{t=2}$ (starting from upper right)



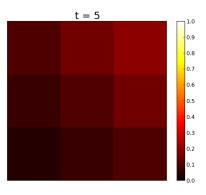
• Distribution $\pi^{t=3}$ (starting from upper right)



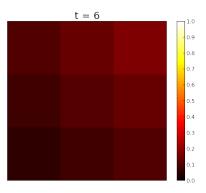
• Distribution $\pi^{t=4}$ (starting from upper right)



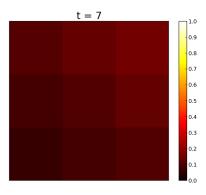
• Distribution $\pi^{t=5}$ (starting from upper right)



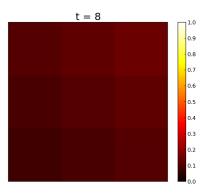
• Distribution $\pi^{t=6}$ (starting from upper right)



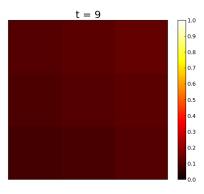
• Distribution $\pi^{t=7}$ (starting from upper right)



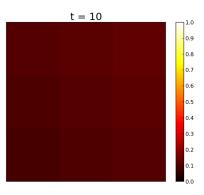
• Distribution $\pi^{t=8}$ (starting from upper right)



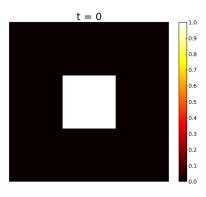
• Distribution $\pi^{t=9}$ (starting from upper right)



• Distribution $\pi^{t=10}$ (starting from upper right)



• Distribution $\pi^{t=0}$ (starting from center)



• Distribution $\pi^{t=1}$ (starting from center)

