# Advanced topics in Markov-chain Monte Carlo

Lecture 4:

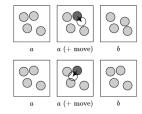
Perfect sampling in Markov-chain Monte Carlo Part 3/3: Coupling from the past / hard spheres

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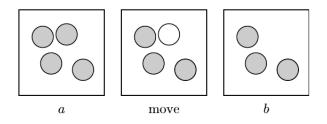
ICFP -Master Course Ecole Normale Supérieure, Paris, France

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#### Hard-sphere simulation (traditional)



#### Hard-sphere simulation (birth-and-death)



$$Z = \sum_{N=0}^{\infty} \lambda^{N} \int \cdots \int dx_{1} \ldots dx_{N} \pi(x_{1}, \ldots, x_{N})$$

- $\pi(a) = \lambda \pi(b)$
- Death probability (per particle, per time interval): 1dt
- Birth probability (per particle, per unit square): λdt

#### Poisson distribution

Poisson distribution (number *n* of events per unit time):

$$\pi_{\Delta t=1}(n)=\frac{\lambda^n \mathrm{e}^{-\lambda}}{n!}$$

Poisson distribution (number n of events per time dt):

$$\pi_{\mathrm{d}t}(n) = \frac{(\lambda \mathrm{d}t)^n \mathrm{e}^{-\lambda \mathrm{d}t}}{n!} \implies \pi_{\mathrm{d}t}(1) = \lambda \mathrm{d}t, \pi_{\mathrm{d}t}(2) = 0$$

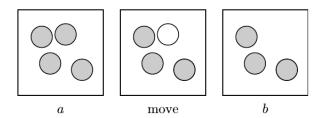
Poisson waiting time: Probability that next event after time t:

$$\mathbb{P}(t) = (1 - \lambda dt), \dots, (1 - \lambda dt)\lambda dt$$

$$\mathbb{P}(t) = \underbrace{\underbrace{(1 - \lambda \mathrm{d}t) \to (1 - \lambda \mathrm{d}t)}_{\mathrm{e}^{-\lambda t}} \lambda \mathrm{d}t}^{\sum \mathrm{d}t = t}$$

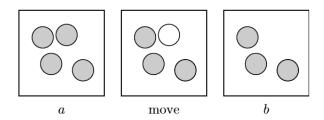
...can be sampled with  $t = (-\log ran[0, 1])/\lambda$ 

#### Birth-and-death (principle 1)



- N spheres, each of them may die.
- a new sphere may be born (but there may be problems).
- rate for next event:  $N + \lambda$ .
- $\mathbb{P}(\text{death}) \propto N$  and  $\mathbb{P}(\text{birth}) \propto \lambda$ , reject if overlap.

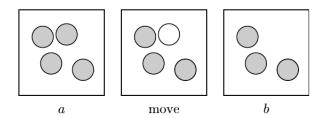
#### Birth-and-death (implementation 1)



- start with N = 0 spheres
- Next-event time :  $-\log ran(0,1)/(N+\lambda)$  (in steps of 1)
- sample random number ran[0, 1]: if smaller than  $\lambda/(\lambda+N)$ : add a disk (reject if overlap), otherwise delete a disk.

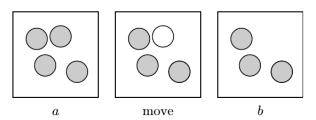
NB: Check configuration at integer time steps, for sampling.

#### Birth-and-death (principle 2)



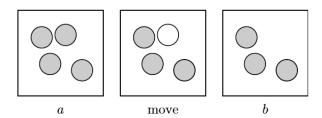
- N spheres, each of them knows when it will die (sad) rate=1.
- a new sphere may be born (but there may be problems) rate  $= \lambda$ .

#### Birth-and-death (implementation 2)



- start with N = 0 spheres.
- Advance to next birth time :  $-\log ran[0,1]/\lambda$  (in steps of 1).
- If no rejection, install death time log ran[0, 1]

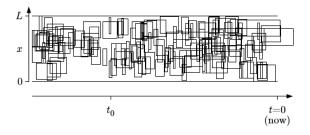
#### Birth-and-death (principle 3)



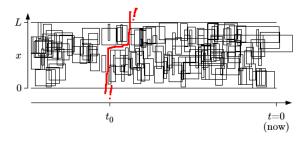
• Hyptothetical spheres are born with rate  $= \lambda$ , and they die with rate 1.

Check later whether all this pans out correctly.

## Birth-and-death (implementation 3)

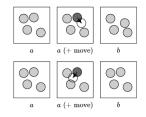


### Birth-and-death (implementation 3)



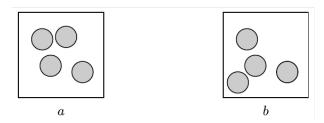
Can be made into a perfect sampling algorithm

#### Hard-sphere simulation (traditional)



Algorithm remains correct if displacement random in box.

#### Path coupling



- Any two configurations of spheres a and z can be connected through a path a → b → c → ..... → z, where any two neighbors differ only in 1 sphere.
- MC algorithm: Take random sphere, place it at random position.
- We can study the probability to go from 1 to 2 differences and the probability to go from 1 to 0 differences (TD).