

## Chapter 5 - Markov chain Monte Carlo

### Power-law distribution

$$Pr(X=k) = \frac{k^s}{\sum_{j=1}^{\infty} j^s} \quad k=1,2,\dots$$

$\Rightarrow$  If sum is finite.

$$Pr(X=k) \propto k^s$$

### Conditional distribution

$X \sim g(x)$  where  $g(x)$  is a pdf

$Y|X \sim t(y|x)$  where  $t(y|x)$  is a pdf

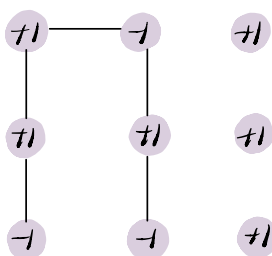
$$h(x|y) = \frac{p(x,y)}{c(y)} \quad t(y|x)g(x)$$

$$c(y) = \int p(x,y) dx = \int t(y|x)g(x) dx$$

$$h(x|y) \propto p(x,y) = t(y|x)g(x)$$

$\uparrow$   
A function of  $X$ .

e.g. Let  $\alpha$  be a configuration of  $\pm 1$  and  $-1$ 's on the vertices of this lattice



$$\text{Energy}(\alpha) = \sum_{i \sim j} -\alpha_i \alpha_j$$

$\approx$  all pairs

$$\text{e.g. } \alpha = (+1, -1, +1, +1, +1, +1, -1, -1, +1)$$

$$\text{Energy}(\alpha) = \# \text{ of pairs with different sign} - \# \text{ of pairs of same sign.}$$

### Gibbs (or Boltzmann) distribution

$$\pi(\alpha) = \frac{e^{-\beta \text{Eng}(\alpha)}}{\sum_{\alpha} e^{-\beta \text{Eng}(\alpha)}}$$

$\sum$  sum of  $2^n$  terms.

$$\beta = 0 \Rightarrow \pi(\alpha) = \frac{1}{2^n} \text{ uniform over possible } \alpha.$$

$$\beta > 0 \Rightarrow \text{favor } \alpha \text{ with more neighbors with the same sign.}$$

$$\beta < 0 \Rightarrow \text{favor } \alpha \text{ with more neighbors with the different sign.}$$

How to generate draws from these distribution? Markov Chain ♥♥

**Borel Distribution**  $\Pr(X=x) = \frac{e^{-\lambda} (\lambda)^x}{x!}$

Goal:  $E(X)$ ,  $\Pr(X \leq c)$

$$X_1, \dots, X_n \stackrel{iid}{\sim} P_X(X)$$

$$\begin{cases} \bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i \\ \hat{q} = \frac{1}{n} \sum_{i=1}^n 1_{X_i \leq c} \end{cases} \quad \text{We know} \quad \begin{cases} \bar{X}_n \xrightarrow{P} E(X_n) \\ \hat{q} \xrightarrow{P} q \end{cases} \quad \begin{cases} E(\bar{X}_n) = E(X_n) \\ E(\hat{q}) = q \end{cases}$$

What if  $X_1 \dots X_n$  not iid? But just  $X_i \sim P_X(X)$ , not independent?

### Strong Law of Large Numbers for Markov chain

Let  $X_0, X_1, \dots, X_n$  be a Markov chain with stationary distribution

$$\pi(x) = P_X(x)$$

Then

$$\frac{r(X_0) + r(X_1) + \dots + r(X_n)}{n+1} \xrightarrow{P} E_{\pi}(r(X)) \stackrel{!}{=} \sum_i r(i) \pi_i$$

for any function  $r(\cdot)$  where  $E_{\pi}(r(X)) < \infty$

Before: Given  $P$ , find  $\pi$ .

Now: Given  $\pi$ , find  $P$ .

Let  $\pi$  be a discrete p.m.f.

$$\pi_k = \Pr(X=k)$$

### Metropolis-Hastings Algorithm. Generate draws from p.m.f

① Choose any irreducible TPM  $T$ , with the same state space as  $\pi$ .

② Choose  $X_0 = i_0$ .

③ For  $n=1, 2, \dots$  Propose to move from  $X_{n-1}=i$  to  $X_n=j$  according to  $T(T_{ij})$

④ Decide to accept the move (set  $X_n=j$ ) with probability

$$a(i,j) = \min\left(1, \frac{\pi_j}{\pi_i} \times \frac{T_{ji}}{T_{ij}}\right)$$

Target  
move if the  
location is better

Proposal  
if you can afford it

Otherwise,  $X_n = i$

Repeat

$\Rightarrow X_0 \dots X_n$  is a Markov chain with TPM  $P$   $\begin{cases} P_{ij} = T_{ij} a(i,j) \text{ for } i \neq j \\ P_{ii} = 1 - \sum_{j \neq i} P_{ij} \text{ for } i=j. \end{cases}$

$P_X(X) = \pi(X)$  will be unique limiting (stationary) distribution of  $P$ .

proof

Let  $P$  be TPM of  $\{X_0, X_1, \dots\}$

For  $i \neq j$ ,  $P_{ij} = \Pr(X_n = j \mid X_{n-1} = i)$

(a) Probability of proposing to move from  $i$  to  $j$  ( $T_{ij}$ )

(b) Probability accepting move ( $a(i, j)$ )

$$\begin{aligned} P_{ij} &= \Pr(\text{probability } i) \Pr(\text{accepting proposal } j) \\ &= T_{ij} a(i, j) \\ &= T_{ij} \min \left\{ 1, \frac{\pi_j}{\pi_i} \frac{T_{ji}}{T_{ij}} \right\} \\ &= \min \left\{ T_{ij}, \frac{\pi_j}{\pi_i} T_{ji} \right\} \end{aligned}$$

$$\text{So, } \pi_i P_{ij} = \min \{ \pi_i T_{ij}, \pi_j T_{ji} \}$$

$$\pi_j P_{ji} = \min \{ \pi_j T_{ji}, \pi_i T_{ij} \}$$

$$\Rightarrow \pi_i P_{ij} = \pi_j P_{ji}$$

$$\Rightarrow P \text{ is time-reversible}$$

$$\Rightarrow \pi \text{ is stationary} \Rightarrow \text{limiting.}$$

$$\text{For } i=j, L_{ii} = 1 - \sum_{j \neq i} P_{ij}$$

Some R

• sample (what, probability, size, replacement=T)