

Markov Chain Monte Carlo

(Digression)

Consider a M.C. $\{X_n, n \geq 1\}$
with trans. probabilities

$$P_{ij} = P(X_k = j \mid X_{k-1} = i) \quad i, j \in S$$

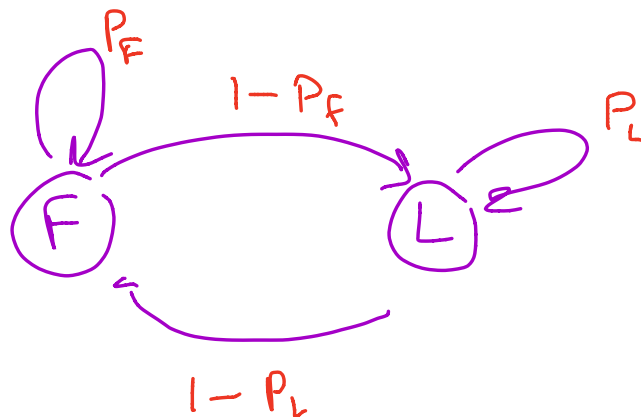
For simplicity assume

$$S = \{1, 2, \dots, n\}$$

discrete &
finite

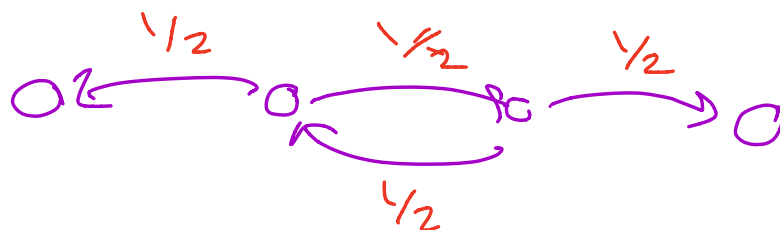
Let $P = (P_{ij})_{1 \leq i, j \leq n}$ be the
p.t.m.

Ex:



$$P = \begin{bmatrix} P_F & 1 - P_F \\ 1 - P_L & P_L \end{bmatrix} \quad S = \{F, L\}$$

Ex:



$$P = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1/2 & 0 & 1/2 & 0 \\ 0 & 1/2 & 0 & 1/2 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Lets compute

$$P(X_{k+l} = j \mid X_k = i) = P_{ij}^{(l)}$$

$$P_{ij}^{(l)} = \sum_{\alpha \in S} P(X_{k+l} = j \mid X_{k+l-1} = \alpha) P(X_{k+l-1} = \alpha \mid X_k = i)$$

(l-1)

$$= \sum_{\alpha \in S} P_{\alpha j} P_{i \alpha}$$

$$P^{(l)} = (P_{ij}^{(l)})_{i,j \in S} \text{ then}$$

$$= P^{(l-1)} P \quad \text{moreover}$$

$$P^{(1)} = P$$

$$P^{(2)} = P^{(1)} P = P^2$$

⋮

$$P^{(m)} = P^m$$


Ergodic Thm: If we have \rightarrow

chain s.t. $\exists n_0$ s.t.

$$\min_{i,j \in S} P_{ij}^{(n_0)} > 0 \quad \text{then}$$

$P_{ij}^{(n)} \mapsto \pi_j > 0$ the stationary dist.

$$P_{ij}^{(n)} = \sum_{\alpha \in S} P_{\alpha j} P_{i\alpha}^{(n-1)}$$



$$\pi_j = \sum_{\alpha \in S} P_{\alpha j} \pi_i$$

$$\Rightarrow \boxed{\pi = P^T \pi}$$

Another version of Ergodic Thrm.

X_1, \dots, X_n is ergodic M.R.

$$\frac{1}{n} \sum_{i=1}^n f(x_i) \rightarrow \mathbb{E} f(x)$$

$$n \sum_{i=1}^n f(x_i) \xrightarrow{\text{a.s.}} \int f(x) d\pi(x)$$

$X \sim \pi$

"time averages converge to space averages"

Suppose we want to sample P to do MC simulation. but we only know P up to a constant.

We could then come up with a MCMC with stat. dist. P .

Use the Ergodic theorem to compute expectations of functions under P .

Metropolis-Hastings

- Target P
- Proposal distribution $Q(x|y)$
- Acceptance Ratio

$$R(\xi^{(t)} | \xi^{(*)}) = \frac{P(\xi^*) Q(\xi^{(t)} | \xi^{(*)})}{P(\xi^{(t)}) Q(\xi^{(*)} | \xi^{(t)})}$$

Alg:

1. Start at $\xi^{(1)}$

2. For $i = 1, 2, \dots, D_0$

(i) Sample candidate

$$\xi^* \sim Q(\cdot | \xi^{(t)})$$

(ii) $p = \min \{1, R(\xi^{(t)}, \xi^{(*)})\}$

$$\text{Let } \mathbf{x}^{(t+1)} = \begin{cases} \mathbf{x}^* & p \\ \mathbf{x}^{(t)} & 1-p \end{cases}$$