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Introduction to the Metropolis Algorithm

Again to get the basic idea we consider the discrete case with

$$p_{ij} > 0$$

Time Reversible Markov Chains

Reverse time order, what is?

$$P\{X_m = j \mid X_{m+1} = i_1, X_{m+2} = i_2, \dots, X_{m+n} = i_n\}$$

??

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$$= \frac{P(X_m = i_1, X_{m+1} = i_2, \dots, X_{m+n} = i_n)}{P(X_{m+1} = i_1, \dots, X_{m+n} = i_n)}$$

$$= P(X_m = i_1) P(X_{m+1} = i_2 | X_m = i_1)$$

$$\frac{P(X_{m+2} = i_3, \dots, X_{m+n} = i_n | X_m = i_1, X_{m+1} = i_2)}{P(X_{m+1} = i_1) P(X_{m+2} = i_2, \dots, X_{m+n} = i_n | X_m = i_1, X_{m+1} = i_2)}$$

$$= \frac{P(X_m = i_1)}{P(X_{m+1} = i_1)} P_{i_1 i_2}$$

$$\equiv \frac{\pi_{i_1} p_{i_1 i_2}}{\pi_{i_2}}$$

$$\text{or } \frac{\pi_{i_2} p_{i_2 i_1}}{\pi_{i_1}}$$

So the reverse process is a Markov chain with

$$P_{ij}^* = \frac{\pi_j p_{ji}}{\pi_i}$$

Time Reversible

A MC is TR if

$$P_{ij}^* = P_{ji}$$

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

"the chance of seeing an
 $i \rightarrow j$ transition

is the same as the chance of
 seeing a
 $j \rightarrow i$ transition"

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Now suppose we have

$$x_i > 0$$

$$\text{s.t. } \sum x_i = 1$$

$$x_i p_{ij} = x_j p_{ji}$$

then

$$\sum x_i p_{ij} = x_j \sum p_{ji} = x_j$$

$$x P = x$$

$$\Rightarrow x = \pi$$

Example: $P = \begin{bmatrix} \frac{1}{4} & \frac{3}{4} \\ \frac{3}{8} & \frac{5}{8} \end{bmatrix}$

$$\pi = \left\{ \frac{1}{3}, \frac{2}{3} \right\}$$

$$\pi_1 P_{12} = \frac{1}{3} \left(\frac{3}{4} \right) = \frac{3}{12}$$

$$\pi_2 P_{21} = \frac{2}{3} \left(\frac{3}{8} \right) = \frac{6}{24} = \frac{3}{12}$$

Example in

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Example

$$P = \begin{bmatrix} .05 & .9 & .05 \\ .05 & .05 & .9 \\ .9 & .05 & .05 \end{bmatrix}$$

$$\pi = \left\{ \frac{1}{3}, \frac{1}{3}, \frac{1}{3} \right\}$$

$$\pi P = \pi$$

$$\pi_1 p_{12} = \frac{1}{3} \left(\frac{9}{10} \right) = \frac{3}{10}$$

$$\pi_2 p_{21} = \frac{1}{3} (.05) \neq \frac{3}{10}.$$

Example

$$\text{suppose } p_{ij} = p_{ji}$$

$$\Rightarrow \pi = \alpha \perp.$$

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Metropolis Hastings Alg

Goal: construct a MC whose stationary distribution is π

know: π_i / π_j & q_{ij}

Let ~~q_{ij}~~ be a transition matrix

Define a new MC based on q_{ij} and π_i / π_j as follows.

Suppose you are at i .

Pick j from q_i .

Compute $\alpha(i, j) = \min\left\{1, \frac{\pi_j q_{ji}}{\pi_i q_{ij}}\right\}$

With prob α go to j .

With prob $1 - \alpha$ stay at i
(repeat ϵ)

Note:

If $j = i$ you repeat

then

$$p_{ij} = q_{ij} \alpha(i,j) \quad i \neq j$$

$$\pi_i p_{ij} = \pi_i q_{ij} \min \left\{ 1, \frac{\pi_j q_{ji}}{\pi_i q_{ij}} \right\}$$

$$= \min \left\{ \pi_i q_{ij}, \pi_j q_{ji} \right\}$$

$$= \pi_j p_{ji}$$

P is reversible with
stationary distribution

$$= \pi$$

Example

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$$Q = \begin{bmatrix} \frac{1}{4} & \frac{3}{4} \\ \frac{3}{4} & \frac{1}{4} \end{bmatrix} \quad \pi_Q = \left[\frac{1}{2}, \frac{1}{2} \right]$$

$$\text{Let } \pi = \left[\frac{1}{3}, \frac{2}{3} \right]$$

$$P_{12} = \frac{3}{4} \min \left\{ 1, \frac{\pi_2}{\pi_1} \right\}$$

$$= \frac{3}{4} \min \left\{ 1, 2 \right\} = \frac{3}{4}$$

$$\Rightarrow P_{11} = \frac{1}{4}$$

$$P_{21} = \frac{3}{4} \min \left\{ 1, \frac{\pi_1}{\pi_2} \right\}$$

$$= \frac{3}{4} \min \left\{ 1, \frac{1}{2} \right\} = \frac{3}{8}$$

$$P = \begin{bmatrix} \frac{1}{4} & \frac{3}{4} \\ \frac{3}{8} & \frac{5}{8} \end{bmatrix}$$

$$\left[\frac{1}{3}, \frac{2}{3} \right] \begin{bmatrix} \frac{1}{4} & \frac{3}{4} \\ \frac{3}{8} & \frac{5}{8} \end{bmatrix} = \left[\frac{1}{12} + \frac{6}{24}, \frac{3}{12} + \frac{10}{24} \right]$$
$$= \left[\frac{1}{3}, \frac{2}{3} \right]$$