



# Markov chain stochastic

model for a sequence of possible events where future events depend only on the current but not on previous events

$$P(X_{n+1} | X_1 = x_1, X_2 = x_2, \dots, X_n = x_n) = P(X_{n+1} | X_n = x_n)$$

if  $P(X_1 = x_1, X_n = x_n) > 0$ ,  $X_1, X_{n+1}$  RVs

## Time homogeneous Markov chain

$$P(X_{h+1}=x | X_h=y) = P(X_h=x | X_{h-1}=y)$$

Stationary distribution  $\Leftrightarrow (X_1 \sim \pi \Rightarrow X_2 | X_1 \sim \pi)$

$$\Leftrightarrow (X_1 \sim \pi \Rightarrow X_h | X_1 \sim \pi)$$

for all  $h$

## irreducible Markov chain

nonzero probability of transitioning (possibly in more than one step) from any state to any other

irreducible Markov chain

$\Rightarrow$  stationary distribution is unique

stationary distribution is usually the limiting distribution of Markov chain

start with  $\theta \sim \pi$ , assume  $\theta \neq \theta^o$

$$\pi(\theta) \underbrace{q(\theta, \theta^o) \propto (\theta, \theta^o)}$$

transition probab. l.t., from  $\theta$  to  $\theta^o$

$$= \pi(\theta) q(\theta, \theta^o) \text{ m.h. } \left( 1, \frac{\pi(\theta^o) q(\theta^o, \theta)}{\pi(\theta) q(\theta, \theta^o)} \right)$$

$$= \text{m.h. } \left( \pi(\theta) q(\theta, \theta^o), \pi(\theta^o) q(\theta^o, \theta) \right)$$

$$= \pi(\theta^o) q(\theta^o, \theta) \propto (\theta^o, \theta)$$

$$\sum_{\theta} \pi(\theta) \bar{q}(\theta, \theta) = \pi(\theta) \underbrace{\sum_{\theta} \bar{q}(\theta, \theta)}_{=1} = \pi(\theta)$$

$$\stackrel{||}{\sum_{\theta} \pi(\theta) q(\theta, \theta)}$$

$\pi$  is invar. and d. distribution

(stationary  $\hat{=}$  invar. and)

# Assignment 1, Exercise 4

$$\begin{pmatrix} X \\ Y \end{pmatrix} \sim \mathcal{N} \left( \begin{pmatrix} m \\ Hm + u \end{pmatrix}, \begin{pmatrix} P & PH' \\ HP & HP H' + R \end{pmatrix} \right)$$

$$\Rightarrow Y \sim \mathcal{N}(Hm + u, HP H' + R)$$

match to equation (2)

$$\pi(\theta) = \frac{p(x|\theta) p(\theta)}{\int p(x|\theta) p(\theta) d\theta} = c p(x|\theta) p(\theta)$$

$$\frac{\pi(\theta^*)}{\pi(\theta)} = \frac{p(x|\theta^*) p(\theta^*)}{p(x|\theta) p(\theta)}$$

$$\begin{aligned} \nabla \log \pi(\theta) &= \nabla \log (c p(x|\theta) p(\theta)) \\ &= \nabla \log(c) + \nabla \log(p(x|\theta) p(\theta)) \\ &= \nabla \log(p(x|\theta) p(\theta)) \end{aligned}$$