

Markov Chain

$x_0, x_1, x_2, x_3, \dots$

Transition probability

$$P(x_i | x_{i-1}, x_{i-2}, \dots, x_1, x_0) = P(x_i | x_{i-1})$$

$$= P(x_i, x_{i-1})$$

(Time homogeneous)

Long term behaviour

$$P(x_i | x_0) = \int dx_{i-1} \int dx_{i-2} \dots \int dx_1 \quad P(x_i, x_{i-1}) \times P(x_{i-1}, x_{i-2}) \\ \times \dots \times P(x_1, x_0)$$

Limit distribution

$$\lambda(x_n) = \lim_{n \rightarrow \infty} P(x_n | x_0)$$

## Stationary distribution

$$x_0 \sim \pi$$

$$x_1 | x_0 \sim p$$

we want  $x_1 \sim \pi \leftarrow$  defines stationary dist  $\pi$

---

$$P(x_0) = \pi(x_0)$$

$$P(x_1) = \int \underbrace{P(x_1 | x_0)}_e \underbrace{P(x_0)}_{\pi} dx_0 =: \pi(x_1)$$

$$\pi(x) = \int dx' \pi(x') p(x', x)$$

condition  
to be stationary

if true then  $\pi$  is stationary dist

the limit dist must be a stationary dist.

Goal is design  $p$  s.t.  $p$  is stationary.  
(target)

### Detailed Balance condition

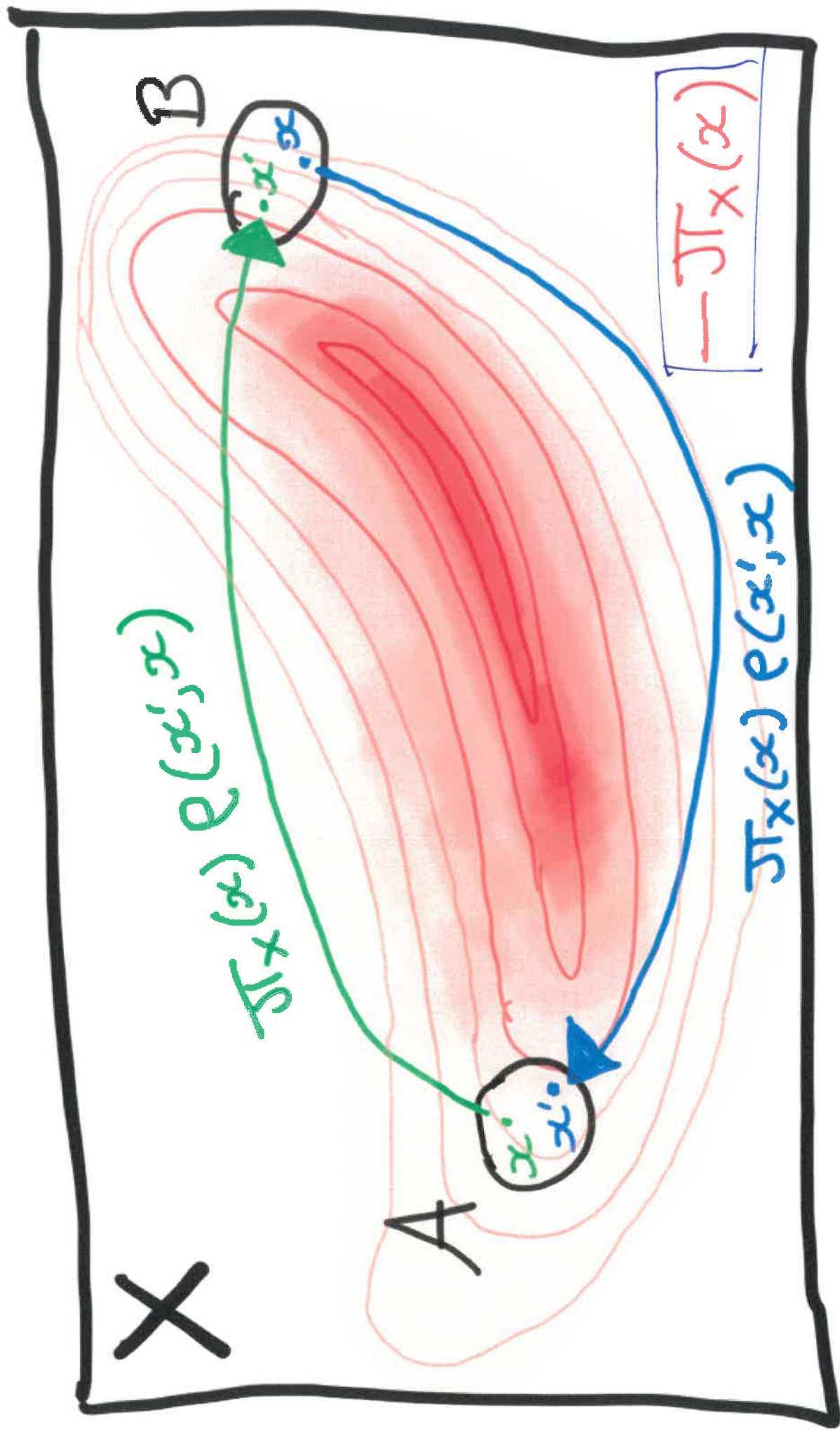
$$\pi(x) p(x', x) = \pi(x') p(x, x')$$

int above DB condition w.r.t.  $x'$  over whole sample space  $X$ .

$$\pi(x) \underbrace{\int dx' p(x', x)}_1 = \int dx' \pi(x') p(x, x')$$

stat condition.

$$\pi(x) = \int dx' \pi(x') p(x, x')$$



two regions  $A, B \subset X$

$$\begin{aligned} P_{\text{res}}(x \in A \text{ and } x' \in B) &= \int_A dx \, \mathcal{I}(x) \int_B dx' \, \rho(x', x) \\ &= \int_A dx \int_B dx' \, \mathcal{I}(x) \, \rho(x', x) \end{aligned}$$

$$\begin{aligned} P_{\text{res}}(x \in B \text{ and } x' \in A) &= \int_B dx \, \mathcal{I}(x) \int_A dx' \, \rho(x', x) \\ &= \int_B dx \int_A dx' \, \mathcal{I}(x) \, \rho(x', x) \\ &= \int_B dx' \int_A dx \, \mathcal{I}(x') \, \rho(x, x') \\ &= \int_A dx' \int_B dx \, \mathcal{I}(x') \, \rho(x, x') \end{aligned}$$

Recall DB condition  $\pi(x) P(x', x) = \pi(x') P(x, x')$

we want  $\pi = P$

MH alg defines  $P(y, x) = \alpha Q(y|x) + (1-\alpha) \sigma^{(d)}(y-x)$   
where  $\alpha = \min\left(1, \frac{P(y) Q(x|y)}{P(x) Q(y|x)}\right)$

$$\alpha(x, y) = \frac{P(x)}{P(y)} \frac{Q(y|x)}{Q(x|y)}$$

$$\alpha(y, x) = 1$$

$$\alpha(x, y) = 1$$
$$\alpha(y, x) = \frac{P(y)}{P(x)} \frac{Q(x|y)}{Q(y|x)}$$

2 cases: ①

②

Consider Case #① LHS DB condition  $P(x) P(y, x) = P(x) Q(y|x) \rightarrow *$

$$\text{RHS DB condition } P(y) P(x, y) = P(y) \left[ \frac{P(x) Q(y|x)}{P(y) Q(x|y)} Q(x|y) + \left(1 - \frac{P(x) Q(y|x)}{P(y) Q(x|y)}\right) \sigma^{(d)}(y-x) \right]$$

2<sup>nd</sup> term is zero.

int w.r.t.  $x$  2<sup>nd</sup> term vanishes

$$P(y)P(x,y) = P(x)Q(y|x) \quad - *$$

MCMC design alg. that

- practical to implement
- only requires the evaluate  $f(x) \propto P(x)$
- produces TH, Markov chain with...
- ...  $P$  that satisfies DB with  $\pi = P$ .

Therefore running for many  $x_n \sim P$  i.e we have  
1 stochastic sample.