

# Markov chain Monte Carlo idea

★ Situation:

- Given a target distribution  $f(x)$
- Want to generate samples from  $f(x)$

★ Idea:

- construct a Markov chain  $\{X_i\}_{i=1}^{\infty}$  so that

$$\lim_{i \rightarrow \infty} P(X_i = x) = f(x)$$

- simulate the Markov chain for many iterations
- for  $m$  large enough  $x_m, x_{m+1}, \dots$  are (essentially) from  $f(x)$

## How to construct the Markov chain

★ How to construct such a Markov chain? ( $x \in \Omega$  discrete)

– Markov chain transition probabilities:

$$P(y|x) = P(X_{i+1} = y | X_i = x)$$

– Need to have

$$f(y) = \sum_{x \in \Omega} f(x)P(y|x) \quad \text{for all } y \in \Omega$$

– Sufficient condition: Detailed balance condition

$$f(x)P(y|x) = f(y)P(x|y) \quad \text{for all } x, y \in \Omega$$

# Metropolis–Hastings

- ★ Detailed balance condition:

$$f(x)P(y|x) = f(y)P(x|y) \quad \text{for all } x, y \in \Omega$$

- ★ Metropolis–Hastings setup for  $P(y|x)$ :

$$P(y|x) = Q(y|x)\alpha(y|x) \quad \text{when } y \neq x$$

$$P(x|x) = 1 - \sum_{y \neq x} Q(y|x)\alpha(y|x)$$

where

- $Q(y|x)$  is a transition matrix (proposal kernel)
- $\alpha(y|x) = \min \left\{ 1, \frac{f(y)}{f(x)} \cdot \frac{Q(x|y)}{Q(y|x)} \right\}$  is an acceptance probability

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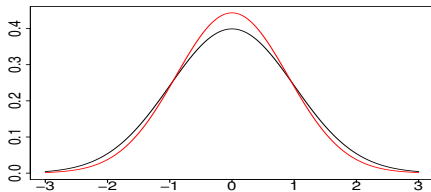
$$P(x|x) = 1 - \sum_{y \neq x} Q(y|x)\alpha(y|x)$$

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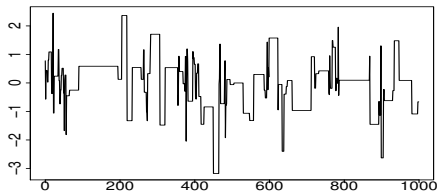
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- ★ We have discussed two examples:
    - toy example: Poisson distribution
    - Ising model:  $x$  is a vector

## Toy example: Independent proposals

- ★ Target distribution:  $x \sim N_{250}(0, I)$
- ★ Proposal distribution:  $y|x \sim N_{250}(0, 0.9^2 \cdot I)$

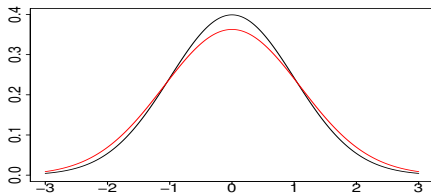


- ★ Trace plot of  $x^1$

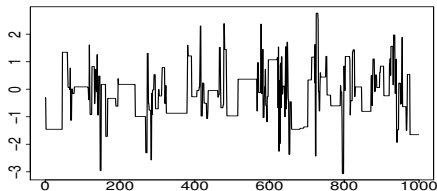


## Toy example: Independent proposals

- ★ Target distribution:  $x \sim N_{250}(0, I)$
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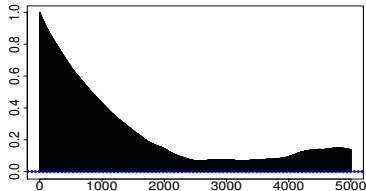
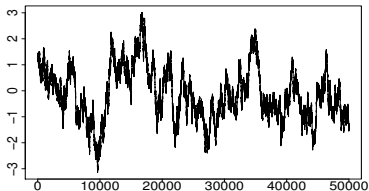
- ★ Trace plot of  $x^1$



# Toy example: Random walk proposals

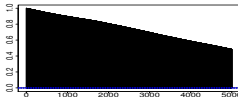
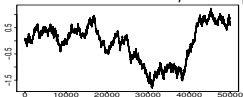
- ★ Target distribution:  $x \sim N_{250}(0, I)$
- ★ Proposal distribution:  $y|x \sim N_{250}(x, \sigma^2 \cdot I)$
- ★ Trace plot and acf of  $x^1$

$\sigma = 0.05$ , acceptance rate = 0.69

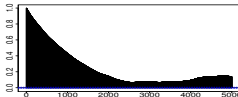
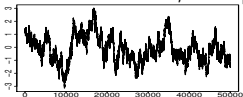


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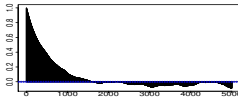
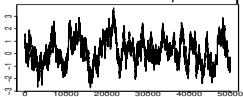
$\sigma = 0.01$ , acceptance rate = 0.94



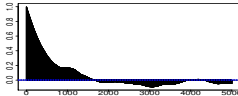
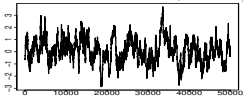
$\sigma = 0.05$ , acceptance rate = 0.69



$\sigma = 0.10$ , acceptance rate = 0.426



$\sigma = 0.20$ , acceptance rate = 0.11



$\sigma = 0.30$ , acceptance rate = 0.018

