

Metropolis-Hastings algorithm

Goal sample from π

We choose a Markov chain with transition probab. l.i.e q Define a new Markov chain which evolves from $\theta = \theta_n$ to θ_{n+1} as follows

- 1) Propose θ^* from $q(\theta, \cdot)$
- 2) Compute $\alpha(\theta, \theta^*) = \min\left(1, \frac{\pi(\theta^*) q(\theta^*, \theta)}{\pi(\theta) q(\theta, \theta^*)}\right)$

3) Set $\theta_{n+1} = \begin{cases} \theta^0 & \text{with prob } \alpha(\theta, \theta^0) \\ \theta & \text{with prob } 1 - \alpha(\theta, \theta^0) \end{cases}$

Only need to know π up to multiplicative constant

Special case Gibbs sampler

Let θ_{-i} be θ with i -th component removed

Suppose we can sample from $\pi(\theta_i | \theta_{-i})$
for all $i \in \{1, \dots, n\}$

Let $\theta^0 = (\theta_1, \theta_1, \dots, \theta_{i-1}, \theta_i^0, \theta_{i+1}, \dots, \theta_n)$

Then $\theta_i^0 = \theta_{-i}$

$$\text{Set } q(\theta, \theta^*) = \pi(\theta^* | \theta_{-1})$$

$$q(\theta^*, \theta) = \pi(\theta_{-1} | \theta^*) = \pi(\theta_{-1} | \theta)$$

$$\frac{\pi(\theta^*) q(\theta^*, \theta)}{\pi(\theta) q(\theta, \theta^*)} = \frac{\pi(\theta^*) \pi(\theta_{-1} | \theta)}{\pi(\theta) \pi(\theta^* | \theta_{-1})}$$

$$= \frac{\pi(\theta^*) \pi(\theta_{-1} | \theta) \pi(\theta_{-1})}{\pi(\theta) \pi(\theta^* | \theta_{-1}) \pi(\theta_{-1})} = \frac{\pi(\theta^*) \pi(\theta)}{\pi(\theta) \pi(\theta^*)} = 1$$

and so $\alpha(\theta, \theta^*) = 1$ We always accept

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(A \cap B) = P(A|B) P(B)$$

partially conjugate pr. or

A pr. or is partially conjugate if the conditional pr. or and conditional posterior belong to the same family of distributions

$$\underbrace{p(\gamma | \theta, \tau)}_{p(\gamma | \theta)} p(\theta | \tau) p(\tau) \sim p(\theta | \tau) p(\tau),$$

const wrt τ