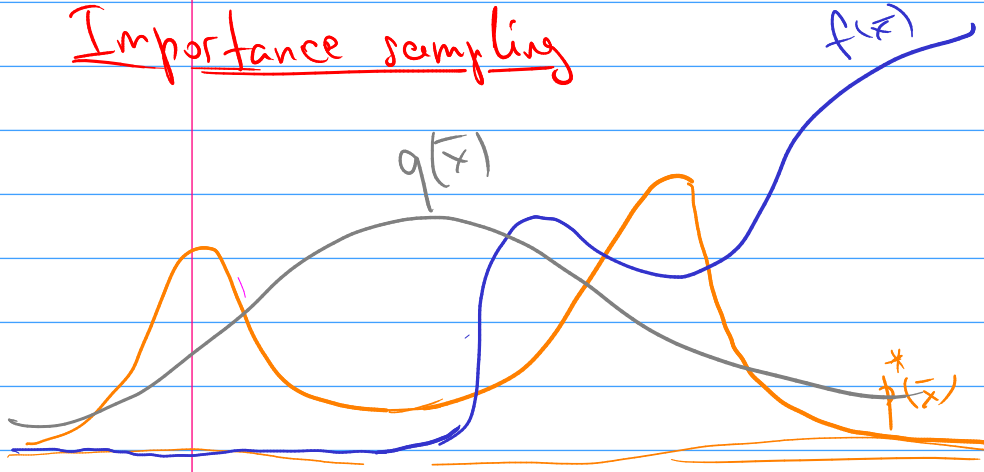


Importance sampling

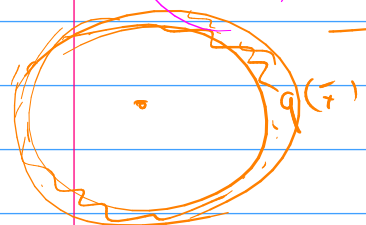
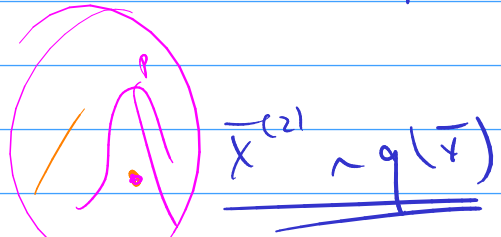


$$E_{p(x)} [f(x)] = \approx \frac{1}{R} \sum_{i=1}^R f(x^{(i)})$$

$$E_{p(x)} [f(x)] = \int f(x) p(x) dx =$$

$$= \int f(x) \frac{p(x)}{q(x)} q(x) dx = E_{q(x)} \left[f(x) \cdot \frac{p(x)}{q(x)} \right] \approx$$

$$\approx \frac{1}{R} \sum_{i=1}^R \left(\frac{p(x^{(i)})}{q(x^{(i)})} f(x^{(i)}) \right)$$



↑ importance weights

$$p(x) = \frac{1}{Z_p} p^*(x)$$

$$q(x) = \frac{1}{Z_q} q^*(x)$$

$$Z_q = \int q^*$$

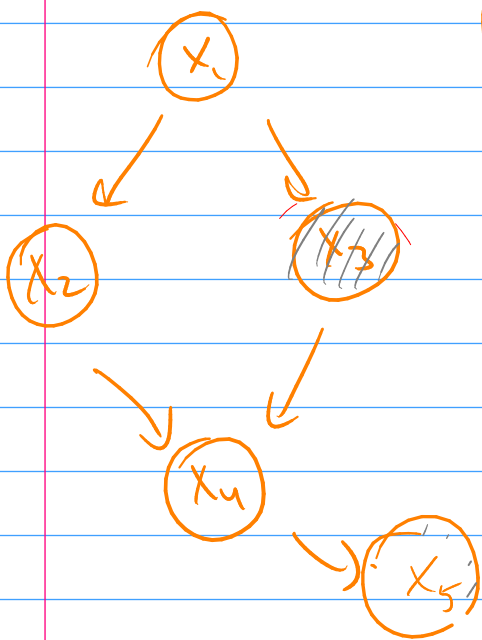
$$E_p [f] = \int p \cdot f dx = \int f \cdot \left(\frac{p}{q} \right) \cdot q dx =$$

$$= \int f \frac{p^*/Z_p}{q^*/Z_q} \cdot q dx \approx \frac{1}{R} \sum \left[\frac{Z_q}{Z_p} \frac{p^*}{q^*} \right] f(x^{(i)})$$

$$\frac{z_p}{z_q} = \int \frac{p^*(\bar{x})}{z_q} d\bar{x} = \int p^*(\bar{x}) \frac{q(\bar{x})}{q^*(\bar{x})} d\bar{x} =$$

$$= \mathbb{E}_q \left[\frac{p^*(\bar{x})}{q^*(\bar{x})} \right] \approx$$

$$\approx \frac{1}{R} \sum_{i=1}^R \frac{p^*(x^i)}{q^*(x^i)}$$



$$p(x_1, \dots, x_5) = p(x_1) p(x_2|x_1) p(x_3|x_1) \cdot$$

$$\cdot p(x_4|x_2, x_3) p(x_5|x_3)$$

$$p(x_1, x_2, x_4, x_5 | \underline{x_3})$$

Rejection sampling:

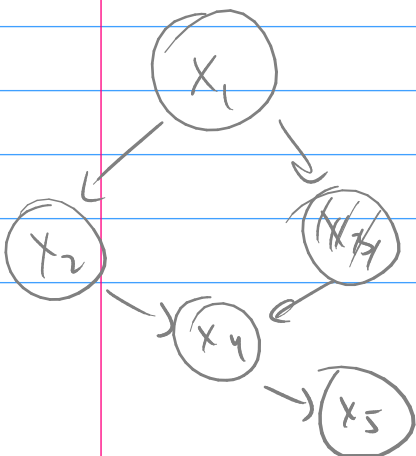
$$\boxed{(x_1, \dots, x_5)} \quad \times$$

Importance sampling

$$\mathbb{E}_p[f(\bar{x})]$$

$$q(\bar{x}) = \underline{x_3} = \text{fix}, \quad x_1 \dots x_5 \sim \text{Unif} \quad q(\bar{x}) = \frac{1}{z}$$

$$\mathbb{E}_q[f(\bar{x})] \approx \frac{1}{R} \sum_{i=1}^R \left[\frac{p(\bar{x}^i)}{q(\bar{x}^i)} \right] \cdot f(\bar{x}^i)$$



$$q: \quad x_1 \sim p(x_1) \quad \underline{x_3 = \text{fix}}$$

$$x_2 \sim p(x_2|x_1)$$

$$x_4 \sim p(x_4|x_2, x_3)$$

$$x_5 \sim p(x_5|x_4)$$

$$(x_1, \dots, x_n)$$

$$p(x_1, \dots, x_5) = p(x_1) p(x_2|x_1) p(x_3|x_2) \dots$$

$$q(x_1, \dots, x_5) = p(x_1) p(x_2|x_1) p(x_4|x_2, x_3) p(x_5|x_4)$$

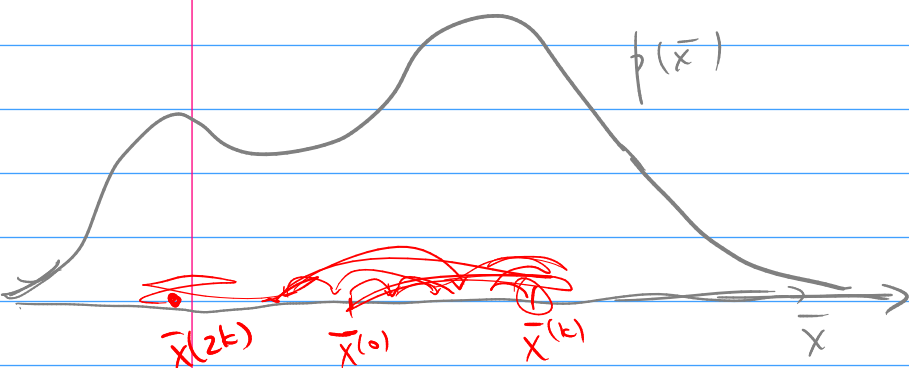
evidence
↓
 $p = p(x|y)$

$$\frac{p(\bar{x})}{q(\bar{x})} = p(x_3|x_2)$$

likelihood weighted sampling

$$E_p[f] \approx \frac{1}{R} \sum_{z=1}^R \left(f(x^{(z)}) \cdot \prod_{x_i \in y} p(x_i | \text{par}(x_i)) \right)$$

MCMC - Markov Chain Monte Carlo



x_1, x_2, \dots

$$p(x_n | x_{n-1}, \dots, x_1) = p(x_n | x_{n-1})$$

$$p^{(1)}(\bar{x}') = \int T(\bar{x}' | \bar{x}) p^{(0)}(\bar{x}) d\bar{x}$$

$$p^{(1)}(\bar{x}) = A p^{(0)}(\bar{x})$$

$$p^{(k)}(\bar{x}) = A p^{(k-1)}(\bar{x})$$

$$p^{(k)}(\bar{x}') = \int T(\bar{x}' | \bar{x}) p^{(k-1)}(\bar{x}) d\bar{x}$$

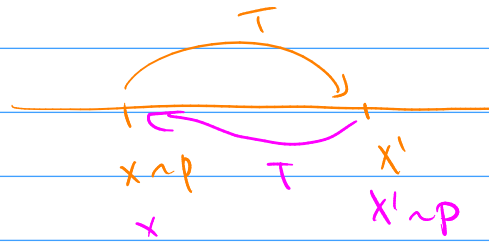
$$p^{(0)}, p^{(1)}, p^{(2)}, \dots$$

$$\rightarrow \underline{\underline{\pi(\bar{x}')}} = \int T(\bar{x}' | \bar{x}) \underline{\underline{\pi(\bar{x})}} d\bar{x}$$

stationary / unique / pers.

Умова
Саманд:

$$\forall x, x' \quad T(x', x) p(x) = T(x, x') p(x')$$



\Rightarrow p - қат. p.

$$\pi(x') \stackrel{?}{=} \int T(x', x) \pi(x) dx = \int \underbrace{T(x, x')}_{\text{сәт.}} \underbrace{\pi(x)}_{\text{сәт.}} dx = \pi(x') \cdot 1$$

Metropolis - Hastings

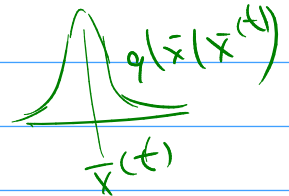
$$\boxed{p(\bar{x})}$$

$q(\bar{x} | \bar{x}^{(t)})$ - sampler

Негізгі үлгі:

$$- \bar{x}' \sim q(\bar{x}' | \bar{x}^{(t)})$$

$$- a(\bar{x}', \bar{x}^{(t)}) = \frac{p(\bar{x}')}{p(\bar{x}^{(t)})} \cdot \frac{q(\bar{x}^{(t)} | \bar{x}')}{q(\bar{x}' | \bar{x}^{(t)})} =$$



$$\boxed{a(\bar{x}', \bar{x}) = \frac{1}{a(\bar{x}, \bar{x}')}}$$

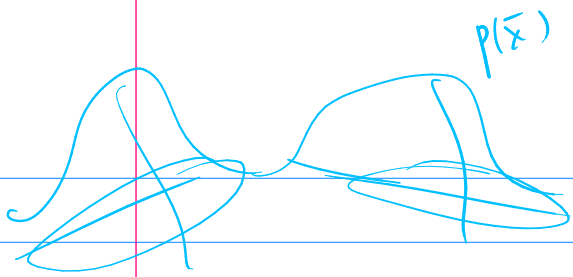
$$= \frac{p^*(\bar{x}')}{p^*(\bar{x}^{(t)})} \cdot \frac{q(\bar{x}^{(t)} | \bar{x}')}{q(\bar{x}' | \bar{x}^{(t)})}$$

- if $a \geq 1$ then $\bar{x}^{(t+1)} = \bar{x}'$

else $\bar{x}^{(t+1)} = \begin{cases} \bar{x}' & \text{сәт. } a \\ \bar{x}^{(t)} & \text{сәт. } 1-a \end{cases}$

$$\underbrace{T(x, x')}_{\text{сәт.}} \underbrace{p(x')}_{\text{сәт.}} \stackrel{?}{=} \underbrace{T(x', x)}_{\text{сәт.}} \underbrace{p(x)}_{\text{сәт.}} \quad \underline{a(\bar{x}, \bar{x}') \leq 1}$$

$$\cancel{q(\bar{x} | \bar{x}')} \cdot \frac{p(\bar{x})}{p(\bar{x}')} \cdot \frac{q(\bar{x}' | \bar{x})}{q(\bar{x} | \bar{x}')} \cdot p(\bar{x}') \stackrel{?}{=} p(\bar{x}) \cdot q(\bar{x}' | \bar{x})$$



$p(\bar{x})$

$$p(\bar{\theta} | D) \propto p(\theta) \prod_n p(\bar{x}_n | \theta)$$

$$\log p^*(\bar{\theta} | D) = \log p(\bar{\theta}) + \sum_n \log p(\bar{x}_n | \bar{\theta})$$