

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

$$\bar{x} - \square$$

$$p^*(\bar{x}) \propto p(\bar{x})$$

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

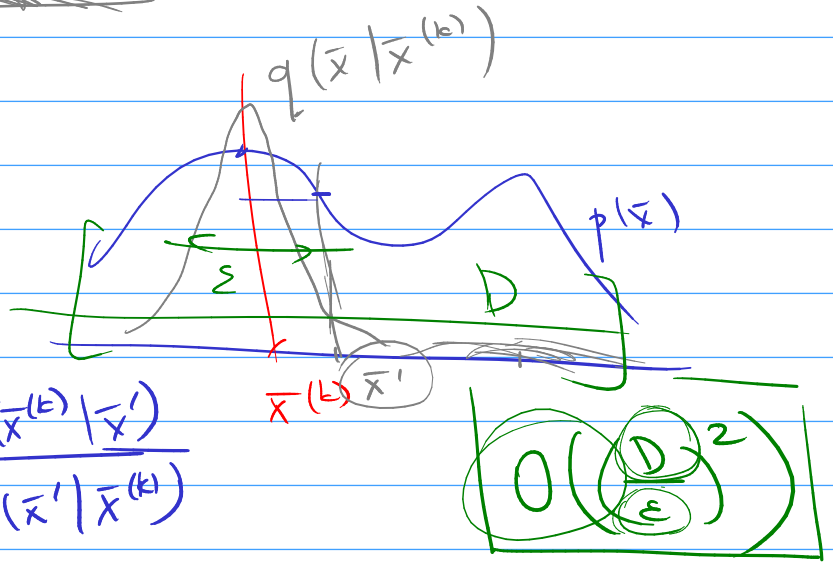
$$n = O(\sqrt{n})$$



MCMC - Metropolis

Metropolis-Hastings

$$p(\bar{x}^{(k)}) \quad \bar{x}' \sim \frac{q(\bar{x}' | \bar{x}^{(k)})}{p(\bar{x}')} p(\bar{x}')$$



$$a(\bar{x}' | \bar{x}^{(k)}) = \frac{p^*(\bar{x}')}{p^*(\bar{x}^{(k)})} \cdot \frac{q(\bar{x}^{(k)} | \bar{x}')}{q(\bar{x}' | \bar{x}^{(k)})}$$

if $a > 1 \Rightarrow$ accept +

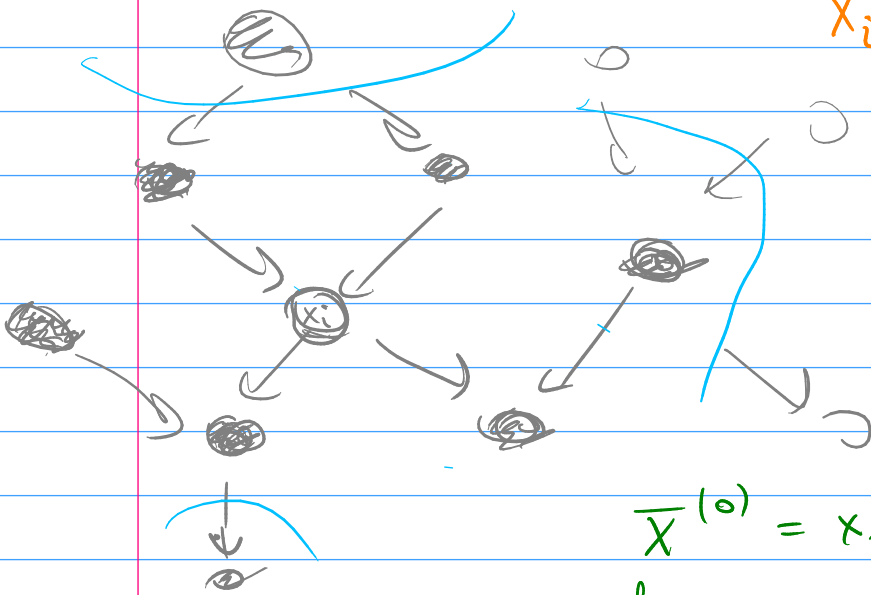
$a < 1 \Rightarrow$ accept $c \cdot \text{log } a$, else $\bar{x}^{(k+1)} = \bar{x}^{(k)}$

Topic modeling

Gibbs sampling

$$p(\bar{x}) = p(x_1, x_2, \dots, x_n)$$

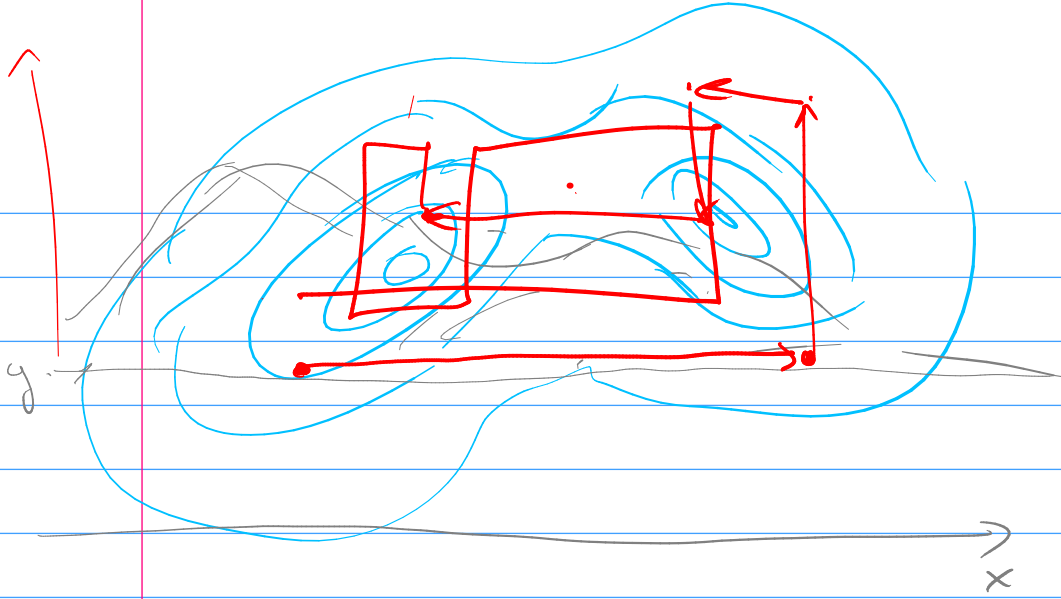
$$x_i \sim p(x_i | x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n) = p(x_i | \bar{x}_{-i})$$



$$\bar{x}^{(0)} = x_1^{(0)}, \dots, x_n^{(0)}$$

- loop:

$$x_i^{(k+1)} \sim p(x_i^{(k+1)} | x_1^{(k+1)}, \dots, x_{i-1}^{(k+1)}, x_{i+1}^{(k)}, \dots, x_n^{(k)})$$



$$q(\bar{x}' | \bar{x}) = \begin{cases} 0, & \bar{x}'_{-i} \neq \bar{x}_{-i} \\ p(x'_i | \bar{x}_{-i}), & \bar{x}'_{-i} = \bar{x}_{-i} \end{cases}$$



$$\bar{x} = (x_i, \bar{x}_{-i})$$

$$\bar{x}' = (x'_i, \bar{x}_{-i})$$

$$\alpha(\bar{x}', \bar{x}) = \frac{p(\bar{x}')}{p(\bar{x})} \cdot \frac{q(\bar{x} | \bar{x}')}{q(\bar{x}' | \bar{x})} =$$

$$= \frac{p(\bar{x}')}{p(\bar{x})} \cdot \frac{p(x_i | \bar{x}_{-i})}{p(x'_i | \bar{x}_{-i})} = \frac{p(x'_i | \bar{x}_{-i}) p(\bar{x}_{-i})}{p(x_i | \bar{x}_{-i}) p(\bar{x}_{-i})} \cdot \frac{p(x_i | \bar{x}_{-i})}{p(x'_i | \bar{x}_{-i})} = 1$$

slice sampling

MCMC

$$\bar{x} \sim p(\bar{x}) \Leftrightarrow (\bar{x}, u) \sim \text{Unif}(\text{support } p(\bar{x}))$$

$$u | \bar{x} = \text{Unif}([0, p(\bar{x})])$$

\bar{x} / u

x_{\min}

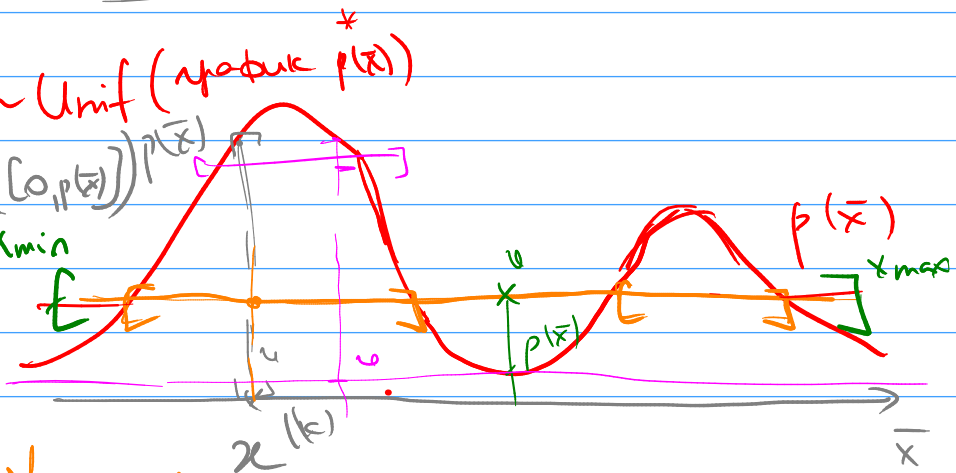
x_{\max}

Algorithm: $x^{(k)} \rightarrow x^{(k+1)}$:

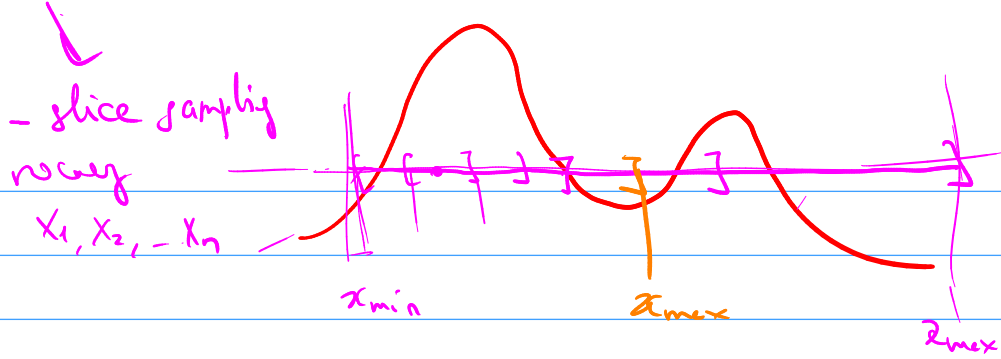
- $u \sim \text{Unif}([0, p(\bar{x}^{(k)})])$

- choose $[x_{\min}, x_{\max}]$ yobereuen

- $x^{(k+1)} \sim \text{Unif}([x_{\min}, x_{\max}] \cap \text{support } p(\bar{x}))$ ypu rou. rejection sampling



- sample $\bar{x}(k)$
 - from samp. slice sampling



Hamiltonian MCMC

(\bar{z}, \bar{z}) phase space
 t
 momentum

$$v_i = \frac{dz_i}{dt}$$

Hamiltonian dynamics

$$p(\bar{z}) = \frac{1}{Z} e^{-E(\bar{z})}$$

$$\frac{dz_i}{dt} = - \frac{\partial E(\bar{z})}{\partial z_i}$$

$$K(\bar{z}) = \frac{1}{2} \|\bar{z}\|^2 = \frac{1}{2} \sum z_i^2$$

$$H(\bar{z}, \bar{z}) = E(\bar{z}) + K(\bar{z})$$

Hamiltonian dynamics

$$\frac{dz_i}{dt} = v_i = \frac{\partial H}{\partial z_i}$$

$$\frac{dH}{dt} = \sum_i \left(\frac{\partial H}{\partial z_i} \frac{dz_i}{dt} + \frac{\partial H}{\partial v_i} \frac{dv_i}{dt} \right) = 0$$

$$\frac{dz_i}{dt} = - \frac{\partial H}{\partial v_i}$$

Tegepa luybura

$$(\bar{z}, \bar{z}) \xrightarrow{t}$$

$$p(\bar{z}, \bar{z}) = \frac{1}{Z_H} e^{-H(\bar{z}, \bar{z})}$$

$$p(\bar{z}) = \frac{1}{Z} e^{-E(\bar{z})}$$



$$\hat{r}_i(t + \frac{\epsilon}{2}) = \hat{r}_i(t) - \frac{\epsilon}{2} \frac{\partial E}{\partial z_i}(\hat{z}(t))$$

$$\hat{z}_i(t + \epsilon) = \hat{z}_i(t) + \epsilon \hat{r}_i(t + \frac{\epsilon}{2})$$

$$\hat{r}_i(t + \frac{3\epsilon}{2}) = \hat{r}_i(t + \frac{\epsilon}{2}) - \epsilon \frac{\partial E}{\partial z_i}(\hat{z}(t + \epsilon))$$

2) bak mana H?

loop!

$$- (\bar{z}, \bar{r})$$

$$= \rightsquigarrow (\bar{z}^{(t)}, \bar{r}^{(t)}) \quad \left. \vphantom{\bar{z}^{(t)}} \right\} H$$

$$p(\bar{z}, \bar{r}) = \frac{1}{Z_H} e^{-H(\bar{z}, \bar{r})}$$

$$- \bar{r} \sim p(\bar{r} | \bar{z}) = \mathcal{N}(\bar{r} | 0, \mathbf{I})$$

$$= \frac{1}{Z_H} e^{-E(\bar{z})} \cdot e^{-K(\bar{r})}$$

$$p(\bar{r} | \bar{z}) \propto e^{-K(\bar{r})} = e^{-\frac{1}{2} \sum r_i^2}$$

$$p(\bar{r} | \bar{z}) = \mathcal{N}(\bar{r} | 0, \mathbf{I})$$

3) Hybrid Monte-Carlo

$$\frac{p(\bar{z}, \bar{r})}{p(\bar{z}', \bar{r}')}$$

$$(\bar{z}, \bar{r}) \xrightarrow{t} (\bar{z}', \bar{r}')$$

$$\left[\begin{array}{l} \text{accept } (\bar{z}', \bar{r}') \text{ c beg-100} \\ \min(1, e^{H(\bar{z}, \bar{r}) - H(\bar{z}', \bar{r}')})) \end{array} \right]$$

PMF

U, V

$$X \approx UV^T$$

$n \times m \quad n \times k \quad m \times k$

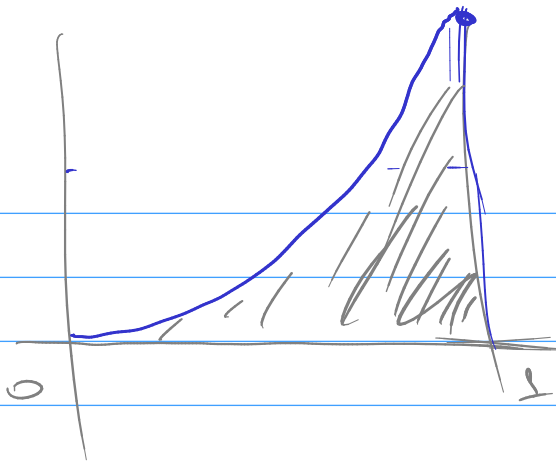
$$p(u) p(v)$$

$$p(u, v | X) \propto p(u) p(v) p(X | u, v)$$

BPMF

$$\left\{ \hat{z}_{i,a} = \frac{1}{50} \sum_{j=1}^{50} \frac{u_i^{(j)} + v_a^{(j)}}{2} \right\} \sim p(u, v | X)$$

3 - Spectra



$$p(x|D) = \mathbb{E}_{p(\theta|D)} [p(x|\theta)]$$

