

これは、観測的に与えられたデータに基づいて

$P(z|y)$ を $P(z)$ と近似する。近似した $P(z|y)$ に
近似した $P(z)$ と

と近似した $P(z)$ と $P(z|y)$ の KL 距離 $D_{KL}(q(z) \parallel p(z|y))$

この最小化問題を解くと $P(z|y)$ は近似した
 $q(z)$ を計算する

結局 D_{KL} を求めるために、

$p(z|y)$ は必要だが無理

↓ 別の方法で

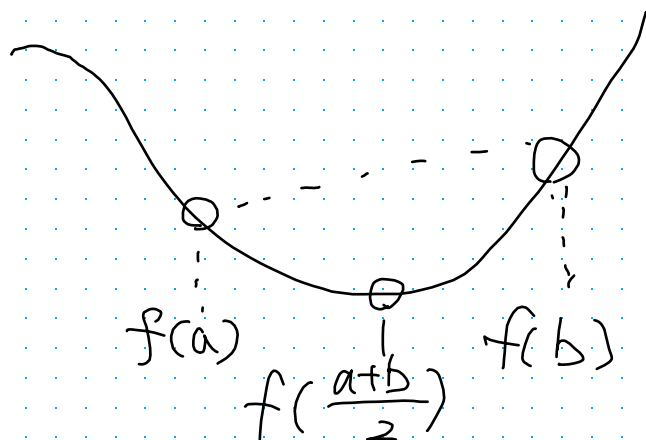
変分 FPE

$$\begin{aligned} \log p(y) &= \log \int \int p(x, z) dz \\ &= \log \int \int q(z) \frac{p(y, z)}{q(z)} dz \\ &\geq \int q(z) \log \frac{p(y, z)}{q(z)} dz = F(q(z)) \end{aligned}$$

↓

Jensen's Inequality:

$$\sum_i \lambda_i f(x_i) \geq f\left(\sum_i \lambda_i x_i\right)$$



$$\geq \int q(z) \log \left[\int q(z) \frac{P(y, z)}{q(z)} dz \right]$$

$$\geq \int q(z) \log \frac{P(y, z)}{q(z)} dz = F(q(x))$$

$$\log P(y) - F[q(z)]$$

$$= \int q(z) \log P(y) dz - \int q(z) \log \frac{P(y|z)}{q(z)} dz$$

$$= \int q(z) \log \frac{P(y) q(z)}{P(y, z)} dz$$

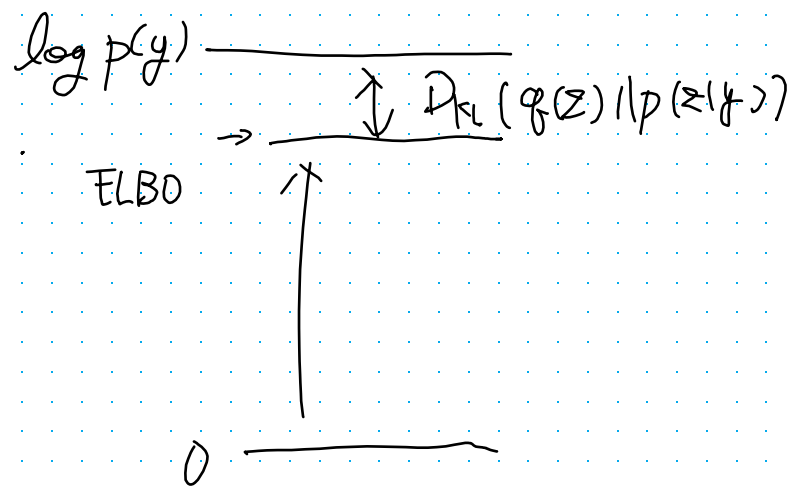
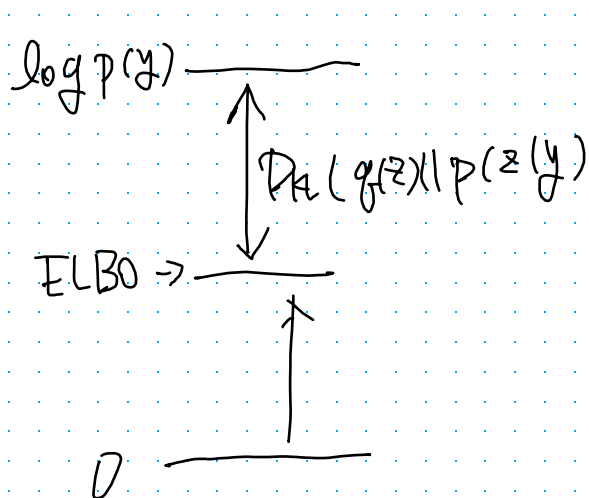
$$= \int q(z) \log \frac{q(z)}{p(z|y)} dz$$

$$= D_{KL}(q(z) \| p(z|y))$$

so, distance between $p(z|y)$ and $q(z)$

is equal to $\log P(y) - F[q(z)]$

対数周辺尤度 = 変分下限 (最大値) の時, KL は最小



so if we want to approximate $p(z|y)$ by using $q(z)$

we just maximize ELBO : $\underbrace{F[q(z)]}$

$$q(z) = \underset{q}{\operatorname{argmax}} F[q(z)]$$

$$= \underset{q}{\operatorname{argmax}} \int q(z) \log \frac{p(x, y)}{q(z)} dz$$

interested in the posterior distribution

$$p(z|x)$$

x : observation

z : latent variables

$$p(z|x) = \frac{p(z, x)}{\int p(z, x) dz} \leftarrow \text{this denominator is not feasible to compute } p(x)$$

VI: variational inference

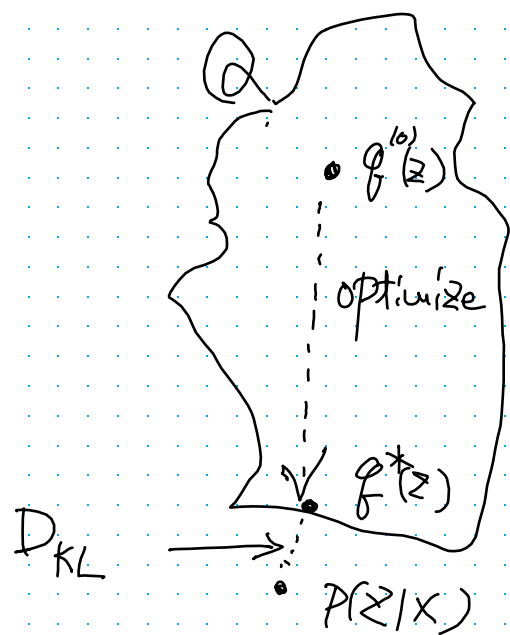
$$\text{Let } q(z) \approx p(z|x)$$

$\uparrow$

How close?

Use D_{KL} as metric

$$q^*(z) = \underset{q(z) \in Q}{\operatorname{argmin}} D_{KL}(q(z) || p(z|x))$$



$$D_{KL}(q(z) \parallel p(z|x))$$

$$= \int q(z) q(z) \log \frac{q(z)}{p(z|x)} dz$$

$$= \mathbb{E}_{q(z)} \left[\log \frac{q(z)}{p(z|x)} \right]$$

$$= \mathbb{E}_{q(z)} \left[\log \frac{q(z)}{\frac{p(z,x)}{p(x)}} \right] = \mathbb{E}_{q(z)} \left[\log \frac{q(z) p(x)}{p(z,x)} \right]$$

$$= \mathbb{E}_{q(z)} [\log q(z)] + \mathbb{E}_{q(z)} [\log p(x)] - \mathbb{E}_{q(z)} [p(z,x)]$$

$$= \underbrace{\mathbb{E}_{q(z)} [\log q(z)] - \mathbb{E}_{q(z)} [\log p(z,x)]}_{-ELBO(q)} - \mathbb{E}_{q(z)} [p(z,x)]$$

Since, we cannot compute the desired D_{KL}

$$= \mathbb{E}_{q(z)} [\log q(z)] - \mathbb{E}_{q(z)} [\log p(z, x)] - \mathbb{E}_{q(z)} [\log p(z, x)]$$

$$- \text{ELBO}(q)$$

$$\text{ELBO}(q) := \mathbb{E}_{q(z)} [\log p(z, x)] - \mathbb{E}_{q(z)} [\log q(z)]$$

then

$$D_{KL}(q(z) \parallel p(z, x)) = p(x) - \text{ELBO}(q)$$

$$p(x) = \text{ELBO}(q) + \underbrace{D_{KL}(q(z) \parallel p(z, x))}_{\geq 0}$$

$$\Leftrightarrow p(x) \geq \text{ELBO}(q)$$

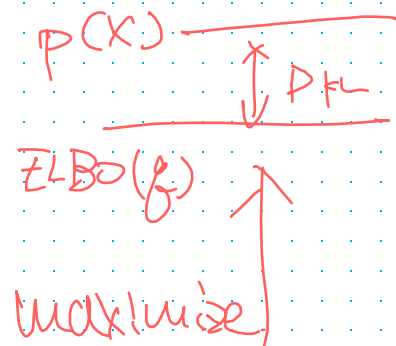
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Since $p(x)$ is fixed quantity.

so if we maximize

$\text{ELBO}(q)$, then it minimize

$$D_{KL}(q(z) \parallel p(z, x))$$



Subject

How to estimate/optimize ELBO

Date

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Attendants

22H_Q3

In deep generative model

gradient-based VI (such as

Variational Auto Encoder is popular

22H_Q3