

$$\log P(z|x) = \log \mathcal{N}(z; \mu, \sigma^2 I)$$

$$\Rightarrow P(z) = \sum_x P(z|x) P(x) = \sum_x \mathcal{N}(0, I) P(x) = \mathcal{N}(0, I) \sum_x P(x) = \mathcal{N}(0, I)$$

$$\Rightarrow \mathcal{L}_{\mu, \sigma^2} = \mathcal{L}_{\mu} + \mathcal{L}_{\sigma^2}$$

$$= \frac{1}{2} \sum_i (\mu_i^2) + \frac{1}{2} \sum_i (\sigma_i^2 - \log \sigma_i^2 - 1)$$

force it

$$\log P(x) = \log \frac{P(x, z)}{P(z|x)} = \log \frac{P(x, z)}{q(z|x)} - \log \frac{P(z|x)}{q(z|x)}$$

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

$$\Rightarrow \log P(x) = \underbrace{\text{ELBO}}_{\text{maximize ELBO} \Rightarrow \text{minimize KL}} + \text{KL}(q(z|x) \| P(z|x)) \Rightarrow \text{as close as possible to } P$$

$$\text{ELBO} = \int q(z|x) \log \frac{P(x, z)}{q(z|x)} dz = \int q(z|x) \log P(x|z) dz + \int q(z|x) \log \frac{P(z)}{q(z|x)} dz$$

$$= \underbrace{\int q(z|x) \log P(x|z) dz}_{(1)} + \underbrace{\int q(z|x) \log \frac{P(z)}{q(z|x)} dz}_{(2)}$$

$$(1) \Rightarrow \text{maximize } \int q(z|x) \log P(x|z) dz$$

$$= \text{maximize } \mathbb{E}_q(\log P(x|z))$$

given  $q(z|x)$ , maximize  $P(x|z)$   
 $\Rightarrow$  LOSS function

$$(2) \Rightarrow -\text{KL}(q(z|x) \| P(z)) = \sum (\exp(\sigma_i^2) - (1 + \sigma_i^2) + \mu_i^2)$$

Appendix B  
 $\leftarrow \ll \text{Auto-encoding Variational Bayes} \gg$

$$\text{sample } z \sim \mathcal{N}(\mu, \sigma^2) = \text{sample } \epsilon \sim \mathcal{N}(0, I)$$

$$z = \mu + \epsilon \times \sigma$$