

1. Making  $P_\theta(x, z)$  tractable with Monte Carlo

2. Making  $P_\theta(x, z)$  tractable with importance sampling

$$\log \sum_z P_\theta(x, z) = \log \sum_z q_\phi(z) \frac{P_\theta(x, z)}{q_\phi(z)}$$

$$= \log E_{z \sim q_\phi(z)} \left[ \frac{P_\theta(x, z)}{q_\phi(z)} \right]$$

$$\approx \log \frac{1}{K} \sum_{k=1}^K \frac{P_\theta(x, z_k)}{q_\phi(z_k)} \quad \text{where } z_k \text{ are sampled from } q_\phi(z_k)$$

Lower variance compared to Monte Carlo.

But what is a good  $q_\phi(z)$ ??

→ Learn  $q_\phi(z)$  from data

Big picture

Our learning objective is to maximise the log probability

$$\log E_{z \sim q_\phi(z)} \left[ \frac{P_\theta(x, z)}{q_\phi(z)} \right] \approx \log \frac{1}{K} \sum_{k=1}^K \frac{P_\theta(x, z_k)}{q_\phi(z_k)}$$