

$$\begin{aligned}
1. \quad \Pi_c(y) &= \underset{x \in C}{\operatorname{argmin}} \|x - y\|^2 \\
&= \underset{(a, b) \in C}{\operatorname{argmin}} (a - y_1)^2 + (b - y_2)^2 \\
&= \begin{cases} (a, 0) & \text{if } y_2 < 0 \\ (a, y_2) & \text{if } 0 \leq y_2 < 1 \\ (a, 1) & \text{if } y_2 \geq 1 \end{cases} \\
&= \begin{bmatrix} a \\ \min \{\max \{y_2, 0\}, 1\} \end{bmatrix}
\end{aligned}$$

□

3.

$$\begin{aligned}
(a) \quad \log(P_\theta(x)) &= \log \mathbb{E}_{z \sim p_z} [P_\theta(x|z)] \\
&= \log \mathbb{E}_{z \sim q_\phi(z|x)} \left[P_\theta(x|z) \frac{p_z(z)}{q_\phi(z|x)} \right] \\
&= \log \frac{1}{K} \sum_{k=1}^K \mathbb{E}_{z \sim q_\phi(z|x)} \left[P_\theta(x|z) \frac{p_z(z)}{q_\phi(z_k|x)} \right] \\
&= \log \frac{1}{K} \sum_{k=1}^K \mathbb{E}_{z_1, \dots, z_K \sim q_\phi(z|x)} \left[P_\theta(x|z_k) \frac{p_z(z_k)}{q_\phi(z_k|x)} \right] \\
&= \log \mathbb{E}_{z_1, \dots, z_K \sim q_\phi(z|x)} \left[\frac{1}{K} \sum_{k=1}^K P_\theta(x|z_k) \frac{p_z(z_k)}{q_\phi(z_k|x)} \right] \\
&\stackrel{\text{Jensen}}{\geq} \mathbb{E}_{z_1, \dots, z_K \sim q_\phi(z|x)} \left[\log \frac{1}{K} \sum_{k=1}^K P_\theta(x|z_k) \frac{p_z(z_k)}{q_\phi(z_k|x)} \right] = VLB_{\theta, \phi}^{(K)}(x)
\end{aligned}$$

$$\begin{aligned}
(b) \quad VLB_{\theta, \phi}^{(K)}(x) &= \mathbb{E}_{z_1, \dots, z_K \sim q_\phi(z|x)} \left[\log \frac{1}{K} \sum_{k=1}^K P_\theta(x|z_k) \frac{p_z(z_k)}{q_\phi(z_k|x)} \right] \\
&= \mathbb{E}_{z_1, \dots, z_K \sim q_\phi(z|x)} \left[\log \mathbb{E}_{I=\{i_1, \dots, i_M\}} \frac{1}{M} \sum_{k=1}^M P_\theta(x|z_{i_k}) \frac{p_z(z_{i_k})}{q_\phi(z_{i_k}|x)} \right] \\
&\stackrel{\text{Jensen}}{\geq} \mathbb{E}_{z_1, \dots, z_K \sim q_\phi(z|x)} \mathbb{E}_{I=\{i_1, \dots, i_M\}} \left[\log \frac{1}{M} \sum_{k=1}^M P_\theta(x|z_{i_k}) \frac{p_z(z_{i_k})}{q_\phi(z_{i_k}|x)} \right] \\
&= \mathbb{E}_{z_1, \dots, z_M \sim q_\phi(z|x)} \left[\log \frac{1}{M} \sum_{k=1}^M P_\theta(x|z_k) \frac{p_z(z_k)}{q_\phi(z_k|x)} \right] = VLB_{\theta, \phi}^{(M)}
\end{aligned}$$

(c) q_ϕ is powerful enough.

$$\Leftrightarrow D_{KL}[q_\phi(\cdot|x_i) \parallel P_\theta(\cdot|x_i)] = 0$$

$$\text{Then, } 0 \leq \log P_\theta(x_i) - VLB_{\theta, \phi}^{(K)} \leq (\log P_\theta(x_i) - VLB_{\theta, \phi}^{(1)}) = D_{KL}[q_\phi(\cdot|x_i) \parallel P_\theta(\cdot|x_i)] = 0$$

$$\therefore \underset{\theta \in \Theta}{\operatorname{maximize}} \sum_{i=1}^N \log P_\theta(x_i) = \underset{\theta \in \Theta, \phi \in \mathcal{G}}{\operatorname{maximize}} \sum_{i=1}^N VLB_{\theta, \phi}^{(K)}(x_i)$$

□

5.

- (a) Define W : event that B win
 L : event that B lose
 D : event that draw

$$\begin{aligned} \mathbb{E}_{P_A, P_B} [\text{points for } B] &= P_{P_A, P_B}(W) - P_{P_A, P_B}(L) \\ &= P(B \text{ plays rock}) P(A \text{ plays scissors}) + P(B \text{ plays paper}) P(A \text{ plays rock}) + P(B \text{ plays scissors}) P(A \text{ plays paper}) \\ &\quad - P(B \text{ plays rock}) P(A \text{ plays paper}) - P(B \text{ plays paper}) P(A \text{ plays scissors}) - P(B \text{ plays scissors}) P(A \text{ plays rock}) \\ &= P(B \text{ plays rock}) (P(A \text{ plays scissors}) - P(A \text{ plays paper})) \\ &\quad + P(B \text{ plays paper}) (P(A \text{ plays rock}) - P(A \text{ plays scissors})) \\ &\quad + P(B \text{ plays scissors}) (P(A \text{ plays paper}) - P(A \text{ plays rock})) \end{aligned}$$

$$\mathbb{E}_{P_A^*, P_B^*} [\text{points for } B] = 0.$$

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$$\therefore \mathbb{E}_{P_A^*, P_B} [\text{points for } B] \leq \mathbb{E}_{P_A^*, P_B^*} [\text{points for } B] \leq \mathbb{E}_{P_A, P_B^*} [\text{points for } B]$$

P_A^*, P_B^* is a solution.

Suppose that one element of solution P_B^{**} is bigger than other.

WLOG, $P(B \text{ plays rock})$ is the biggest. Clearly, $P_A^{**} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$.

$$\text{Then, } \mathbb{E}_{P_A^{**}, P_B^{**}} [\text{points for } B] \leq \mathbb{E}_{P_A^*, P_B} [\text{points for } B]$$

$$\text{where } \tilde{P}_B := \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

$\therefore P_A^{**}, P_B^{**}$ is not a solution in this case.

$$\therefore P_A^* = \begin{bmatrix} 1/3 \\ 1/3 \\ 1/3 \end{bmatrix}, P_B^* = \begin{bmatrix} 1/3 \\ 1/3 \\ 1/3 \end{bmatrix} \text{ is the unique solution.}$$

(b) If $P_B = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$,

$$P(A \text{ win} | A \text{ plays rock}) = P(B \text{ plays scissors} | A \text{ plays rock}) = P(B \text{ plays scissors}) = \frac{1}{3}$$

$$\text{Likewise, } P(A \text{ win} | A \text{ plays paper}) = P(A \text{ win} | A \text{ plays scissors}) = \frac{1}{3}$$

$$\begin{aligned} P(A \text{ win}) &= P(A \text{ win} | A \text{ plays rock}) P(A \text{ plays rock}) \\ &\quad + P(A \text{ win} | A \text{ plays paper}) P(A \text{ plays paper}) \\ &\quad + P(A \text{ win} | A \text{ plays scissors}) P(A \text{ plays scissors}) \end{aligned}$$

It means any strategy is optimal for player A.

