

Siddhant Mishra-Sharma (MIT/AI FI) Summer School

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KL-divergence



A measure of similarity between two probability distributions

$$D_{\text{KL}}(Q||P) = \int_{-\infty}^{\infty} \mathrm{d}x \, q(x) \log \left(\frac{q(x)}{p(x)} \right)$$

Not symmetric!!

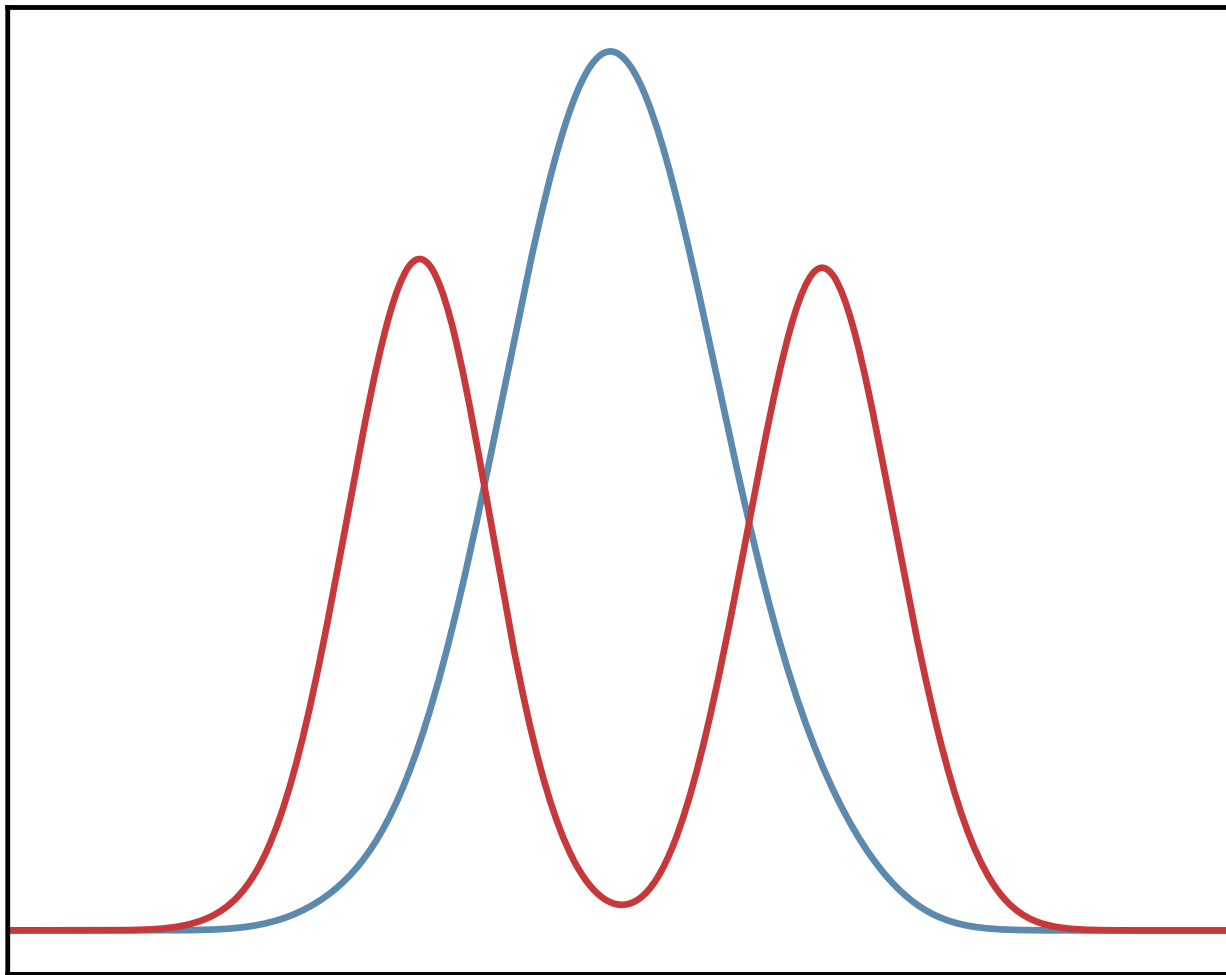
$$D_{KL}(Q||P) \neq D_{KL}(P||Q)$$

Maximum-likelihood inference is equivalent
to minimizing the *forward* KL

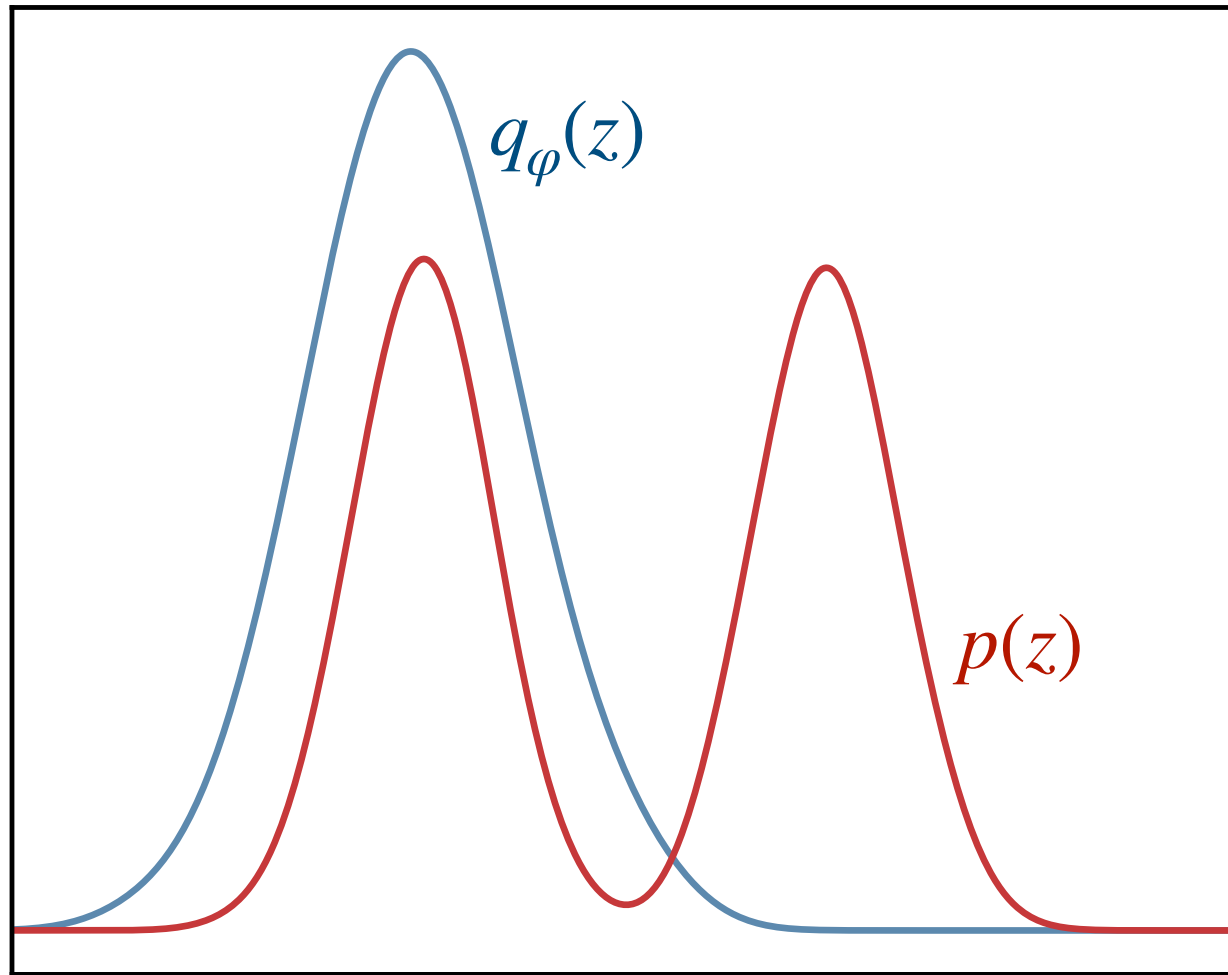
Forward KL

$$D_{\text{KL}}(P_{\mathcal{D}} \parallel Q_{\varphi}) = - \left\langle \log q_{\varphi}(z) \right\rangle_{z \sim p_{\mathcal{D}}(z)} + \text{const} .$$

“Forward” KL $D_{\text{KL}}(P||Q)$



“Reverse” KL $D_{\text{KL}}(Q||P)$





True's distribution

Non-negative! $D_{\text{KL}}(Q||P) \geq 0$

KL-divergence

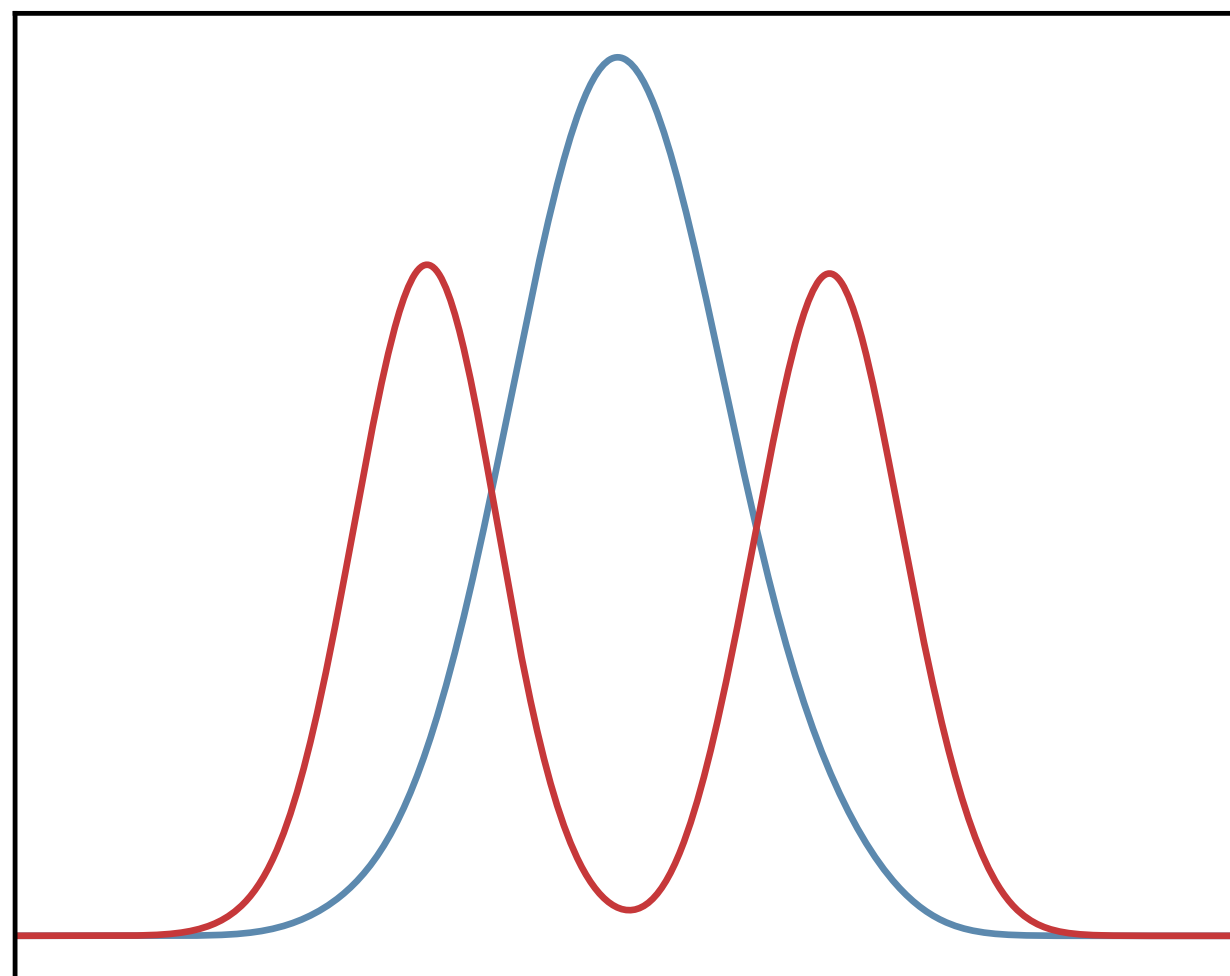
A measure of similarity between two probability distributions

Not symmetric! $D_{\text{KL}}(Q\|P) \neq D_{\text{KL}}(P\|Q)$

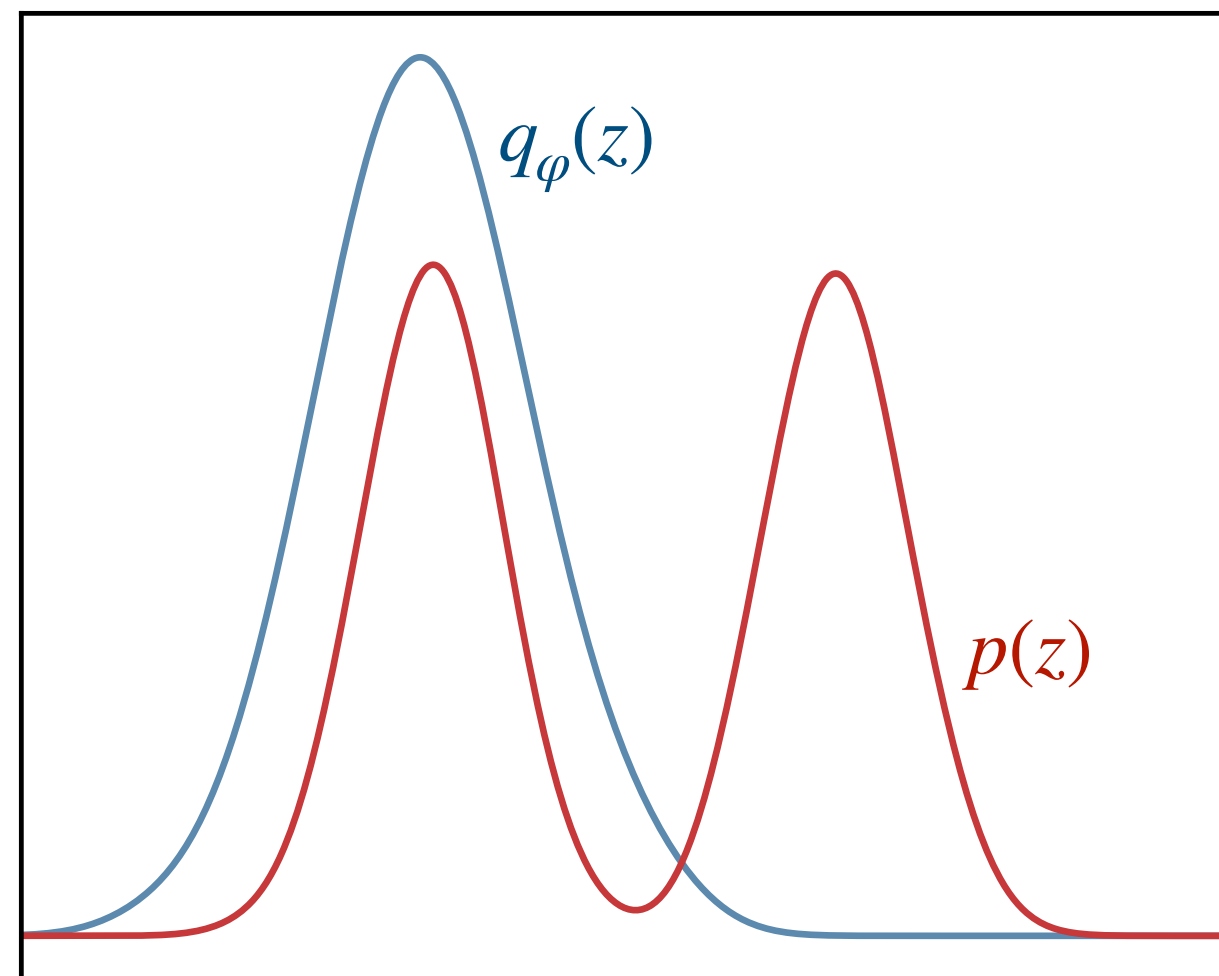
"True" distribution



"Forward" KL $D_{\text{KL}}(P\|Q)$



"Reverse" KL $D_{\text{KL}}(Q\|P)$



$$D_{\text{KL}}(Q\|P) = \int_{-\infty}^{\infty} dx \, q(x) \log \left(\frac{q(x)}{p(x)} \right)$$

Forward KL

$$D_{\text{KL}}(P_{\mathcal{D}}\|Q_{\phi}) = - \left\langle \log q_{\phi}(z) \right\rangle_{z \sim p_{\mathcal{D}}(z)} + \text{const.}$$

Maximum-likelihood inference is equivalent to minimizing the *forward* KL

Non-negative! $D_{\text{KL}}(Q\|P) \geq 0$

Variational inference

Infer the posterior over the latent parameters