Machine Learning Homework: Week 7 & 8

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Statement

Prove the following lemma.

Lemma 1. For any $\delta > 0$,

$$\mathbb{P}_{S \sim D^n} \left(\mathbb{E}_{h \sim \mathcal{P}} \left[e^{n(err_D(h) - err_S(h))^2} \right] \geqslant 3/\delta \right) \leqslant \delta$$

Solution

First we prove that for some fixed $h \sim \mathcal{P}$,

$$\mathbb{E}_{S \sim D^n} \left[e^{n(err_D(h) - err_S(h))^2} \right] \leqslant 3$$

Denote $|err_D(h) - err_S(h)|$ by $\Delta(h)$, by applying Chernoff bound,

$$\mathbb{P}_{S \sim D^n}(\Delta(h) \geqslant \varepsilon) \leqslant 2 \exp(-2n\varepsilon^2)$$

Thus

$$\mathbb{E}_{S \sim D^n} \left[e^{n\Delta(h)^2} \right] = \int_0^{+\infty} \mathbb{P}_{S \sim D^n} \left(e^{n\Delta(h)^2} \geqslant t \right) dt$$

$$= \int_1^{+\infty} \mathbb{P}_{S \sim D^n} \left(\Delta(h) \geqslant \sqrt{\frac{\ln t}{n}} \right) dt + 1$$

$$\leqslant \int_1^{+\infty} 2e^{-2\ln t} dt + 1$$

$$= 3$$

Then by applying Markov Inequality we get

$$\mathbb{P}_{S \sim D^n} \left(\mathbb{E}_{h \sim \mathcal{P}} [e^{n\Delta(h)^2}] \geqslant 3/\delta \right) \leqslant \frac{\mathbb{E}_{S \sim D^n} \left(\mathbb{E}_{h \sim \mathcal{P}} [e^{n\Delta(h)^2}] \right)}{3/\delta} = \frac{\mathbb{E}_{h \sim \mathcal{P}} \left(\mathbb{E}_{S \sim D^n} [e^{n\Delta(h)^2}] \right)}{3/\delta} \leqslant \frac{\mathbb{E}_{h \sim \mathcal{P}} \left(3 \right)}{3/\delta} = \delta$$

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Statement

Determine $D_{KL}(\mathcal{Q}||\mathcal{P})$ for $\mathcal{P} = \mathcal{N}(\mathbf{0}, I_d)$ and $\mathcal{Q} = \mathcal{N}(\mu \mathbf{w}, I_d)$, where $\mathbf{w} \in \mathbb{R}^d$ satisfies $||\mathbf{w}||^2 = 1$, μ stands for the scale factor.

Solution

$$D_{KL}(Q||\mathcal{P}) = \int_{\mathbb{R}^d} \frac{1}{(2\pi)^{d/2}} \exp\left[-\frac{1}{2} \|\mathbf{x} - \mu\mathbf{w}\|^2\right] \frac{1}{2} \left(\|\mathbf{x}\|^2 - \|\mathbf{x} - \mu\mathbf{w}\|^2\right) d\mathbf{x}$$

$$= \int_{\lambda} \int_{\mathbf{y} \in \mathbb{R}^{d-1}, \mathbf{y} \perp \mathbf{w}} \frac{1}{(2\pi)^{d/2}} \exp\left[-\frac{1}{2} \|\lambda\mathbf{w} + \mathbf{y} - \mu\mathbf{w}\|^2\right] \frac{1}{2} \left(\|\lambda\mathbf{w} + \mathbf{y}\|^2 - \|\lambda\mathbf{w} + \mathbf{y} - \mu\mathbf{w}\|^2\right) d\lambda d\mathbf{y}$$

$$= \int_{\lambda} \int_{\mathbf{y} \in \mathbb{R}^{d-1}, \mathbf{y} \perp \mathbf{w}} \frac{1}{(2\pi)^{d/2}} \exp\left[-\frac{1}{2} (\lambda - \mu)^2 - \frac{1}{2} \|\mathbf{y}\|^2\right] \frac{1}{2} \left(\lambda^2 + \|\mathbf{y}\|^2 - (\lambda - \mu)^2 - \|\mathbf{y}\|^2\right) d\lambda d\mathbf{y}$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left[-\frac{1}{2} (\lambda - \mu)^2\right] \frac{1}{2} (2\lambda\mu - \mu^2) d\lambda \left[\int_{\mathbf{y} \in \mathbb{R}^{d-1}, \mathbf{y} \perp \mathbf{w}} \frac{1}{(2\pi)^{(d-1)/2}} \exp\left(-\frac{1}{2} \|\mathbf{y}\|^2\right) d\mathbf{y}\right]$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left[-\frac{1}{2} (\lambda - \mu)^2\right] (\lambda\mu - \mu^2) d\lambda + \frac{\mu^2}{2}$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left[-\frac{1}{2} (\lambda - \mu)^2\right] \mu d\frac{(\lambda - \mu)^2}{2} + \frac{\mu^2}{2}$$

$$= \frac{\mu^2}{2}$$