

Probset #2

3.(b)

Show that MAP estimation with a zero-mean Gaussian prior over θ , specifically $\theta \sim \mathcal{N}(0, \eta^2 I)$, is equivalent to applying L_2 regularization with MLE estimation.

Show that: $\theta_{\text{MAP}} = \underset{\theta}{\operatorname{argmin}} \{-\log p(y|x, \theta) + \lambda \|\theta\|_2^2\}$

$$\theta_{\text{MAP}} = \underset{\theta}{\operatorname{argmax}} p(\theta|x, y)$$

$$= \underset{\theta}{\operatorname{argmax}} p(y|\theta, x) p(\theta)$$

prior: $\theta \sim \mathcal{N}(0, \eta^2 I)$

$$p(\theta) = \frac{1}{\eta \sqrt{2\pi}} e^{-\frac{1}{2}(\frac{\theta}{\eta})^2}$$

$$\rightarrow \underset{\theta}{\operatorname{argmax}} p(y|\theta, x) \times \frac{1}{\eta \sqrt{2\pi}} \exp(-\frac{1}{2}(\frac{\theta}{\eta})^2)$$

$$= \underset{\theta}{\operatorname{argmax}} \log \left\{ p(y|\theta, x) \times \frac{1}{\eta \sqrt{2\pi}} \exp(-\frac{1}{2}(\frac{\theta}{\eta})^2) \right\}$$

$$= \underset{\theta}{\operatorname{argmin}} - \left\{ \log p(y|\theta, x) + \log \frac{1}{\eta \sqrt{2\pi}} + \log \exp(-\frac{1}{2}(\frac{\theta}{\eta})^2) \right\}$$

$$= \underset{\theta}{\operatorname{argmin}} - \left\{ \log p(y|\theta, x) - \frac{1}{2}(\frac{\theta}{\eta})^2 \right\}$$

$$= \underset{\theta}{\operatorname{argmin}} \left\{ -\log p(y|\theta, x) + \frac{1}{2}(\frac{\theta}{\eta})^2 \right\}$$

$$\therefore \lambda = \frac{1}{2\eta^2}$$