

1.2 $\theta = [\theta_0, \dots, \theta_1, \dots, \theta_p]^T$

$$p(\theta) = \frac{1}{\sqrt{|2\pi\Sigma_0|}} \exp\left(-\frac{1}{2}(\theta - \mu_0)^T \Sigma_0^{-1} (\theta - \mu_0)\right)$$

Σ_0 prior covariance

μ_0 prior mean

Set $\Sigma_0 = I, \mu_0 = 0$

$$\rightarrow p(\theta) = \frac{1}{\sqrt{|2\pi I|}} \exp\left(-\frac{1}{2} \theta^T \theta\right)$$

a) $p(\theta|y) = \frac{p(y|\theta) p(\theta)}{p(y)} \propto p(y|\theta) p(\theta)$

$$p(y|\theta) = \left(\frac{1}{\sqrt{2\pi}\sigma}\right)^N \prod_{i=1}^N \exp\left(-\frac{(y_i - (\theta_0 + \theta_1 x_{i1} + \dots + \theta_p x_{ip}))^2}{2\sigma^2}\right)$$

$$p(\theta|y) = \left(\frac{1}{\sqrt{2\pi}\sigma}\right)^N \prod_{i=1}^N \exp\left(-\frac{(y_i - (\theta_0 + \theta_1 x_{i1} + \dots + \theta_p x_{ip}))^2}{2\sigma^2}\right) \frac{1}{\sqrt{|2\pi I|}} \exp\left(-\frac{1}{2} \theta^T \theta\right)$$

$$\log(p(\theta|y)) = -N \log(\sqrt{2\pi}\sigma) - \sum_{i=1}^N \frac{(y_i - (\theta_0 + \theta_1 x_{i1} + \dots + \theta_p x_{ip}))^2}{2\sigma^2} - \frac{1}{2} \theta^T \theta - \log(\sqrt{|2\pi I|})$$

b) derive MAP estimate $\rightarrow \frac{\partial \log(\dots)}{\partial \theta} \stackrel{!}{=} 0$

$$\frac{\partial \log(p(\theta|y))}{\partial \theta} = \frac{\partial}{\partial \theta} \left(\underbrace{-N \log(\sqrt{2\pi}\sigma)}_{\text{vanish}} - \frac{1}{2} \theta^T \theta - \underbrace{\log(\sqrt{|2\pi I|})}_{\text{vanish}} - \sum_{i=1}^N \frac{(y_i - (\theta_0 + \theta_1 x_{i1} + \dots + \theta_p x_{ip}))^2}{2\sigma^2} \right)$$

$y \in \mathbb{R}^N$

$\theta \in \mathbb{R}^p$

$X \in \mathbb{R}^{N \times p}$

$$= \frac{\partial}{\partial \theta} \left(-\frac{1}{2} \theta^T \theta - \frac{1}{2\sigma^2} \|y - X\theta\|^2 \right)$$

$$= -\theta - \frac{1}{2\sigma^2} \frac{\partial}{\partial \theta} ((y - X\theta)^T (y - X\theta))$$

$$= -\theta - \frac{1}{2\sigma^2} (y^T y - y^T X\theta - \theta^T X^T y + \theta^T X^T X \theta)$$

$$= -\theta - \frac{1}{2\sigma^2} (-2X^T y + 2X^T X \theta) = -\theta - \frac{1}{\sigma^2} (-X^T y + X^T X \theta)$$

$$\stackrel{!}{=} 0$$

$$-\theta - \frac{1}{\sigma^2} (-X^T Y + X^T X \theta) = 0$$

$$-\theta \left(I + \frac{1}{\sigma^2} X^T X \right) + \frac{1}{\sigma^2} X^T Y = 0$$

$$\Rightarrow \hat{\theta} = \left(X^T X + \sigma^2 I \right)^{-1} X^T Y$$

← shift of σ^2