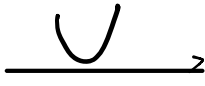
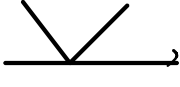
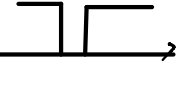


第十一章 笔记

误差	$\varepsilon = \theta - \hat{\theta}$
代价函数	$C(\varepsilon)$
贝叶斯风险	$R = E(C(\varepsilon))$
类型	<div> 2次  $p(\theta x)$ 的均值 </div> <div> 绝对  $p(\theta x)$ 的中值 </div> <div> 成功-失败  $p(\theta x)$ 的众数 </div>

绝对值误差代价函数

$$g(\hat{\theta}) = \int |\theta - \hat{\theta}| p(\theta|x) d\theta$$

$$= \int_{-\infty}^{\hat{\theta}} (\hat{\theta} - \theta) p(\theta|x) d\theta + \int_{\hat{\theta}}^{+\infty} (\theta - \hat{\theta}) p(\theta|x) d\theta$$

$$\text{又有 } \frac{\partial}{\partial u} \int_{\phi_1(u)}^{\phi_2(u)} h(u, v) dv$$

$$= \int_{\phi_1(u)}^{\phi_2(u)} \frac{\partial h(u, v)}{\partial u} dv + \frac{d\phi_2(u)}{du} h(u, \phi_2(u)) - \frac{d\phi_1(u)}{du} h(u, \phi_1(u))$$

$$\text{所以 } \frac{dg(\hat{\theta})}{d\hat{\theta}} = \int_{-\infty}^{\hat{\theta}} p(\theta|x) d\theta - \int_{\hat{\theta}}^{+\infty} p(\theta|x) d\theta$$

$$\text{令其为0则 } \int_{-\infty}^{\hat{\theta}} p(\theta|x) d\theta = \int_{\hat{\theta}}^{+\infty} p(\theta|x) d\theta$$

$$\text{即 } \hat{\theta} \text{ 为后验 } p \text{ 的中值 } \Leftrightarrow P(\theta \leq \hat{\theta}|x) = \frac{1}{2} \text{ 或 } \frac{1}{2}$$

成功-失败型函数

$$g(\hat{\theta}) = \int_{-\infty}^{\hat{\theta}-\delta} 1 \cdot p(\theta|x) d\theta + \int_{\hat{\theta}+\delta}^{\infty} 1 \cdot p(\theta|x) d\theta$$

$$= 1 - \int_{\hat{\theta}-\delta}^{\hat{\theta}+\delta} p(\theta|x) d\theta$$

对给定的 δ 选择 $\hat{\theta}$ 为使 g 的函数

向量函数为标量

$$p(\theta|x) = \frac{p(x|\theta)p(\theta)}{\int p(x|\theta)p(\theta) d\theta}$$

$$\hat{\theta}_1 = E(\theta_1|x) = \int \theta_1 p(\theta_1|x) d\theta_1$$

$$\hat{\theta}_i = E(\theta_i|x) = \int \theta_i p(\theta_i|x) d\theta_i$$

$$E[(\theta_i - \hat{\theta}_i)^2] = \int (\theta_i - \hat{\theta}_i)^2 p(x, \theta_i) dx d\theta_i$$

$$\text{而 } p(\theta_i|x) = \int \dots \int p(\theta|x) d\theta_1 \dots d\theta_p$$

$$\int \theta_i p(\theta_i|x) d\theta_i = \int \theta_i \left[\int \dots \int p(\theta|x) d\theta_1 \dots d\theta_p \right] d\theta_i$$

$$= \int \theta_i p(\theta|x) d\theta$$

$$\text{所以 } \hat{\theta} = \begin{bmatrix} \int \theta_1 p(\theta|x) d\theta \\ \int \theta_2 p(\theta|x) d\theta \\ \vdots \\ \int \theta_n p(\theta|x) d\theta \end{bmatrix} = \int \theta p(\theta|x) d\theta$$

最大后验估计是 MAP

$$\hat{\theta} = \arg \max_{\theta} p(\theta|x)$$

$$= \arg \max_{\theta} P(x|\theta) P(\theta)$$

$$= \arg \max_{\theta} [\ln P(x|\theta) + \ln P(\theta)]$$