

Stat 301

$$X \sim P_\theta \quad (P_\theta : \theta \in \Theta)$$

$$\mathcal{L}(\hat{\theta}, \theta) \quad R(\hat{\theta}, \theta) = \bar{E}_\theta \mathcal{L}(\hat{\theta}, \theta) = \int \mathcal{L}(\theta(x), \theta) dP_\theta(x).$$

e.g.  $X_1, \dots, X_n$  iid  $N(\mu, \sigma^2)$ .  $\mu \in \mathbb{R}$ . loss  $(\hat{\mu} - \mu)^2$

$$Q: \text{Is } \bar{X} \text{ minimax?} \quad R(\bar{X}, \mu) = \bar{E}_\theta (\bar{X} - \mu)^2 = \frac{\sigma^2}{n}$$

Consider  $\pi = N(0, \tau^2)$ .

$$-\frac{\mu^2}{2\tau^2} - \frac{1}{2} \sum_{i=1}^n \frac{(x_i - \mu)^2}{\sigma^2}$$

$$\pi(\mu | X) \propto \pi(\mu) \prod_{i=1}^n p(x_i | \mu) \propto e^{-f(\mu)}$$

$$f(\mu) = \frac{\mu^2}{\tau^2} + \sum_{i=1}^n \frac{(x_i - \mu)^2}{\sigma^2} \quad f'(\mu) = \frac{2\mu}{\tau^2} + \frac{1}{\sigma^2} \sum_{i=1}^n 2(\mu - x_i) = 0,$$

$$\Leftrightarrow \frac{\mu}{\tau^2} + \frac{n\mu}{\sigma^2} = \frac{1}{\sigma^2} \sum_{i=1}^n x_i$$

$$\Rightarrow \hat{\mu} = \mathbb{E}(\hat{\mu}|X) = \frac{\frac{1}{\sigma^2} \sum_{i=1}^n X_i}{\frac{1}{\sigma^2} + \frac{n}{\sigma^2}} = \frac{n/\sigma^2}{\frac{1}{\sigma^2} + n/\sigma^2} \bar{X}$$

$$\begin{aligned} R(\hat{\mu}, \mu) &= \text{Var}(\hat{\mu}) + (\mathbb{E}\hat{\mu} - \mu)^2 \\ &= \left( \frac{n/\sigma^2}{\frac{1}{\sigma^2} + n/\sigma^2} \right)^2 \frac{\sigma^2}{n} + \left( \frac{1/\sigma^2}{\frac{1}{\sigma^2} + n/\sigma^2} \right) \mu^2. \end{aligned}$$

$$\begin{aligned} \int R(\hat{\mu}, \mu) \pi(\mu) d\mu &= \left( \frac{n/\sigma^2}{\frac{1}{\sigma^2} + n/\sigma^2} \right)^2 \frac{\sigma^2}{n} + \left( \frac{1/\sigma^2}{\frac{1}{\sigma^2} + n/\sigma^2} \right)^2 \mu^2 \\ &= \frac{1}{\frac{1}{\sigma^2} + \frac{n}{\sigma^2}}. \end{aligned}$$

to prove  $\bar{X}$  is minimax.  $\sup_{\mu \in \mathbb{R}} R(\bar{X}, \mu) = \frac{\sigma^2}{n}$ .

$$\forall \hat{u}, \quad \sup_{u \in R} R(\hat{u}, u) \geq \int R(\hat{u}, u) \pi(u) du \xrightarrow{\sim} N(0, \tau^2)$$

$$\geq \inf_{\hat{u}} \int R(\hat{u}, u) \pi(u) du.$$

$$= \frac{1}{\frac{1}{\tau^2} + \frac{n}{\sigma^2}}$$

Letting  $\tau^2 \rightarrow \infty$  from both sides, we get

$$\lim_{\tau^2 \rightarrow \infty} \sup_{u \in R} R(\hat{u}, u) \geq \lim_{\tau^2 \rightarrow \infty} \frac{1}{\frac{1}{\tau^2} + \frac{n}{\sigma^2}}$$

$$\Rightarrow \sup_{u \in R} R(\hat{u}, u) \geq \frac{\sigma^2}{n} = R(\bar{X}, u)$$

□

Theorem:  $\exists \{\pi_m\}$  s.t.

$$\sup_{\theta \in \Theta} R(\hat{\theta}, \theta) = \lim_{m \rightarrow \infty} \inf_{\hat{\theta}} \int R(\hat{\theta}, \theta) \pi_m(\theta) d\theta.$$

then  $\hat{\theta}$  is minimax.

Admissibility:

Def:  $\hat{\theta}$  is inadmissible, if  $\exists \tilde{\theta}$  st.

$$\textcircled{1} \quad R(\tilde{\theta}, \theta) \leq R(\hat{\theta}, \theta) \quad \forall \theta \in \Theta$$

$$\textcircled{2} \quad R(\tilde{\theta}, \theta_0) < R(\hat{\theta}, \theta_0) \quad \text{for some } \theta_0.$$

$\hat{\theta}$  is admissible if it is not inadmissible.

Theorem: If  $\hat{\theta}$  is Bayes, then it is admissible.

Proof: Suppose  $\hat{\theta}$  is inadmissible.  $\exists \tilde{\theta} \text{ s.t.}$

$$\textcircled{1} \quad R(\tilde{\theta}, \theta) \leq R(\hat{\theta}, \theta) \quad \forall \theta \in \Theta$$

$$\textcircled{2} \quad R(\tilde{\theta}, \theta) < R(\hat{\theta}, \theta_0) \quad \text{for some } \theta_0 \in \Theta.$$

$\Rightarrow \exists$  an open set  $\Theta_0 \ni \theta_0$  and  $\varepsilon > 0$ ,

s.t.  $R(\tilde{\theta}, \theta) \leq R(\hat{\theta}, \theta) - \varepsilon \quad \forall \theta \in \Theta_0$ .

$$\Rightarrow \int R(\tilde{\theta}, \theta) \pi(\theta) d\theta = \int_{\Theta_0} R(\tilde{\theta}, \theta) \pi(\theta) d\theta + \int_{\Theta_0^c} R(\tilde{\theta}, \theta) \pi(\theta) d\theta$$

$$< \int_{\Theta_0} R(\hat{\theta}, \theta) \pi(\theta) d\theta + \int_{\Theta_0^c} R(\hat{\theta}, \theta) \pi(\theta) d\theta.$$

$$= \int R(\hat{\theta}, \theta) \pi(\theta) d\theta$$

$\Rightarrow \hat{\theta}$  is not Bayes.

□

Complete class theorem (Brown, 1986)

$\hat{\theta}$  is admissible  $\Rightarrow \exists \{\pi_m\}$  s.t.  $\hat{\theta}_{\pi_m} \rightarrow \hat{\theta}$ .

e.g.  $X_1, \dots, X_n \sim N(\theta, 1) \quad \theta \in \mathbb{R} \quad (\hat{\theta} - \theta)^2$ .

Q: Is  $\bar{X}$  admissible? Yes. Wald (1939).

Proof (Blyth's method): Suppose  $\hat{\theta} = \bar{X}$  is inadmissible.

$\exists \tilde{\theta}$  s.t.  $\left\{ \begin{array}{l} \text{① } R(\tilde{\theta}, \theta) \leq \frac{1}{n} \quad \forall \theta \in \mathbb{R}, \\ \text{② } R(\tilde{\theta}, \theta) < \frac{1}{n} \quad \exists \theta_0 \in \mathbb{R}. \end{array} \right.$

$\exists a < b, \varepsilon > 0$  s.t.  $(a, b) \ni \theta_0 \quad \& \quad R(\tilde{\theta}, \theta) < \frac{1}{n} - \varepsilon, \forall \theta \in (a, b)$

consider  $\pi_m = N(0, m)$ .  $\int R(\hat{\theta}_{\pi_m}, \theta) \pi(\theta) d\theta = \frac{1}{h + \frac{1}{m}}$

$$\frac{1}{h} - \int R(\tilde{\theta}, \theta) \pi_m(\theta) d\theta = \frac{1}{h} - \frac{1}{h + \frac{1}{m}} = \frac{1}{h} \cdot \frac{\frac{1}{m}}{h + \frac{1}{m}} \asymp \frac{1}{m}$$

$$\frac{1}{h} - \int R(\tilde{\theta}, \theta) \pi_m(\theta) d\theta \xrightarrow{\text{split}} \int \frac{1}{h} \cdot \pi_m(\theta) d\theta = \int_{(c,b)} + \int_{(a,b)^c}$$

$$= \frac{1}{h} - \int_{(a,b)} R(\tilde{\theta}, \theta) \pi_m(\theta) d\theta - \int_{(a,b)^c} R(\tilde{\theta}, \theta) \pi_m(\theta) d\theta$$

$$= \int_{(a,b)} \left( \frac{1}{h} - R(\tilde{\theta}, \theta) \right) \pi_m(\theta) d\theta + \int_{(a,b)^c} \left( \frac{1}{h} - R(\tilde{\theta}, \theta) \right) \pi_m(\theta) d\theta$$

$$\geq \varepsilon \int_{(a,b)} \pi_m(\theta) d\theta = \varepsilon P(a < N(0, m) < b)$$

$$= \varepsilon P\left(\frac{a}{\sqrt{m}} < N(0, 1) < \frac{b}{\sqrt{m}}\right).$$

$$= \varepsilon \int_{a/\sqrt{m}}^{b/\sqrt{m}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \asymp \frac{1}{\sqrt{m}}.$$

$\exists m$  sufficiently large s.t.

$$\frac{1}{n} - \int R(\tilde{\theta}, \theta) \pi_m(\theta) d\theta > \frac{1}{n} - \int R(\hat{\theta}_{\pi_m}, \theta) \pi_m(\theta) d\theta$$

$$\Leftarrow \int R(\tilde{\theta}, \theta) \pi_m(\theta) d\theta < \int R(\hat{\theta}_{\pi_m}, \theta) \pi_m(\theta) d\theta -$$

$\Rightarrow$  impossible.  $\Rightarrow \bar{X}$  is admissible

□

e.g.  $X_1, \dots, X_n$  iid  $N(\theta, I_2)$        $\|\hat{\theta} - \theta\|^2$   
 $\theta \in \mathbb{R}^2$

Q: Is  $\bar{X}$  admissible?

yes (proved by Stein).

e.g.  $X_1, \dots, X_n$  iid  $N(\theta, I_3)$        $\|\hat{\theta} - \theta\|^2$   
 $\theta \in \mathbb{R}^3$

Q: Is  $\bar{X}$  admissible?

no. James - Stein estimator

$$\hat{\theta}_{JS} = \left(1 - \frac{p-2}{n\|\bar{X}\|^2}\right) \bar{X}$$

Theorem:  $E_\theta \|\hat{\theta}_{JS} - \theta\|^2 < E_\theta \|\bar{X} - \theta\|^2 \quad \forall \theta \in \mathbb{R}^p$   
 for  $p \geq 3$ .

(Brown, 1971).