

Math 281B HW5.

1.  $X \sim \text{Bin}(n, p)$ , and  $\hat{p} = \frac{\sqrt{n}}{1+\sqrt{n}} \left( \frac{X}{n} \right) + \frac{1}{1+\sqrt{n}} \cdot \frac{1}{2}$ , the minimax for  $p$ .  
 prove that  $\sqrt{n}(\hat{p} - p) \xrightarrow{d} N\left(\frac{1}{2} - p, p(1-p)\right)$ .

$$\begin{aligned} \sqrt{n}(\hat{p} - p) &= \sqrt{n} \left( \frac{\sqrt{n}}{1+\sqrt{n}} \left( \frac{X}{n} \right) + \frac{1}{1+\sqrt{n}} \cdot \frac{1}{2} - p \right) \\ &= \frac{\sqrt{n}}{1+\sqrt{n}} \cdot \underbrace{\sqrt{n} \left( \frac{X}{n} - p \right)}_{(I)} + \frac{\sqrt{n}}{1+\sqrt{n}} \underbrace{\left( \frac{1}{2} - p \right)}_{(II)} \end{aligned}$$

Since  $\sqrt{n} \left( \frac{X}{n} - p \right) \xrightarrow{d} N(0, p(1-p))$  by CLT,

$\frac{\sqrt{n}}{1+\sqrt{n}} \rightarrow 1$ . By Slutsky's theorem,

$(I) \xrightarrow{d} N(0, p(1-p))$ ,  $(II) \rightarrow \left(\frac{1}{2} - p\right)$ .

Therefore

$$\sqrt{n}(\hat{p} - p) \xrightarrow{d} N\left(\frac{1}{2} - p, p(1-p)\right).$$

$$2. \quad R_{\frac{x}{n}}(p) = \frac{1}{n} p(1-p) \quad R_{\hat{p}}(p) = \frac{1}{4(1+n)^2}$$

Determine  $I_n = \{p: R_{\hat{p}}(p) \leq R_{\frac{x}{n}}(p)\}$  and ~~its~~ its behavior when  $n \rightarrow \infty$ .

~~Set~~  ~~$R_{\frac{x}{n}}(p)$~~  If  $p \neq \frac{1}{2}$ ,  $p(1-p) < \frac{1}{4}$ , then

$$\frac{R_{\frac{x}{n}}(p)}{R_{\hat{p}}(p)} = \frac{p(1-p)}{1/4} \cdot \frac{(1+n)^2}{n} \xrightarrow{n \rightarrow \infty} 4p(1-p) < 1.$$

Therefore, eventually  $R_{\frac{x}{n}}(p) < R_{\hat{p}}(p)$ .

$$\text{If } p = \frac{1}{2}, \quad R_{\frac{x}{n}}(p) = R_{\frac{x}{n}}\left(\frac{1}{2}\right) = \frac{1}{4n} > R_{\hat{p}}\left(\frac{1}{2}\right) = \frac{1}{4(1+n)^2}$$

To sum up:  $I_n \rightarrow \left\{\frac{1}{2}\right\}$ .

3.  $X \sim \text{Bin}(n, p)$ ,  $L(p, d) = \frac{(d-p)^2}{p(1-p)}$ . Show that:

1.  $\frac{X}{n}$  has constant risk

2.  $\frac{X}{n}$  is the Bayes solution with respect to uniform prior.

1.  $R_{\frac{X}{n}}(p) = E\left(L(p, \frac{X}{n})\right) = E\left(\frac{(\frac{X}{n} - p)^2}{p(1-p)}\right) = \frac{1}{n}.$

2. we have  $\pi(p) = 1$  and  $f(x|p) = \binom{n}{x} p^x (1-p)^{n-x}$

therefore  $\pi(p|x) \propto p^x (1-p)^{n-x}$

which indicates that the posterior is  $\text{Beta}(x+1, n-x+1)$

To minimize the risk, which is equivalent to

$$\begin{aligned} & \min_d \int_0^1 \text{Beta}(x+1, n-x+1) \frac{(d-p)^2}{p(1-p)} p^x (1-p)^{n-x} dp \\ &= \min_d \int_0^1 \text{Beta}(x+1, n-x+1) (d-p)^2 p^{x-1} (1-p)^{n-x-1} dp. \quad (*) \end{aligned}$$

(i)  $x > 0$ ,  $x < n$ : (\*) is equivalent to minimize squared loss under  $\text{Beta}(x, n-x)$ , which is the mean of such distribution:  $\frac{x}{n}$ .

(ii)  $x=0$ : It's easy to see that any  $d$  such that  $d \neq 0$  on a positive ~~measure~~ measure set will result in  $(*) = +\infty$ .

Therefore  $d=0 = \frac{x}{n}$ .

(iii)  $x=n$ : Similar to (ii),  $d=1 = \frac{x}{n}$ .

To sum up:  $\frac{X}{n}$  is the ~~posterior~~ Bayes estimator.

4. Suppose  $\delta$  is unbiased for  $g(\theta)$ . Then there is  $c \in (0, 1)$  such that  $c\delta$  dominates  $\delta$ .

By the bias-variance relationship:

$$\begin{aligned} R_{c\delta}(\theta) &= \text{var}(c\delta) + \text{Bias}^2(c\delta) \\ &= c^2 \text{var}(\delta) + (1-c)^2 g^2(\theta) \\ &= c^2 (R_\delta(\theta) + g^2(\theta)) - 2g^2(\theta)c + g^2(\theta) \end{aligned}$$

This is minimized by  $c = \frac{g^2(\theta)}{R_\delta(\theta) + g^2(\theta)}$

Since  $c=1$  is a root for equation  $R_{c\delta}(\theta) = R_\delta(\theta)$  for all  $\theta$ , then for given  $\theta$ , any number in  $(\frac{g^2(\theta)}{R_\delta(\theta) + g^2(\theta)}, 1)$  will make  $R_{c\delta}(\theta) < R_\delta(\theta)$ .

To make such a choice independent of  $\theta$ , just take

$$\begin{aligned} c &= \bigcap_{\theta \in \Theta} \left( \frac{g^2(\theta)}{R_\delta(\theta) + g^2(\theta)}, 1 \right) \\ &= \left( \sup_{\theta} \frac{g^2(\theta)}{R_\delta(\theta) + g^2(\theta)}, 1 \right). \end{aligned}$$

5.  $X \sim \text{pois}(\lambda)$ . for which  $(a, b)$  such that  $aX + b$  admissible?

First:  $R_{\delta}(X) = \text{var}(\delta) + \text{bias}^2(\delta) = a^2\lambda + ((a-1)\lambda + b)^2$

We consider a prior  $\text{Gamma}(\alpha, \beta)$  on  $\lambda$  (shape-rate parameterization)

it's easy to see that:

$$\pi(\lambda|X) \propto \pi(\lambda) \cdot f(X|\lambda) \propto \lambda^{\alpha+X-1} e^{-(\beta+1)\lambda}$$

So the posterior is  $\text{Gamma}(\alpha+X, \beta+1)$ , with Bayes estimator

$$\frac{\alpha+X}{\beta+1} = \frac{1}{\beta+1} X + \frac{\alpha}{\beta+1}. \quad \text{This is admissible, and by taking}$$

different  $(\alpha, \beta)$ , we see that  $a < 1$ ,  $b > 0$  makes  $aX + b$  admissible.

Now we rule out other options. First,  $(a, b) > 0$ , otherwise it's dominated by  $\delta^+ = \sqrt{\lambda} \mathbb{1}(\sqrt{\lambda} > 0)$ . since  $\lambda > 0$ .

when  $a \geq 1$ .  $R_{\delta}(\frac{\lambda}{a}) \geq a^2\lambda \geq \lambda = R_X(\lambda)$ , so it's dominated by  $X$ .

The case  $\delta = X$  is not admissible by question (4), since  $X$  is unbiased to  $\lambda$ .