

Review : Conditional Expectation

$$\underline{E[X|Y=y]} = \int x f(x|y) dx = g(y)$$

Iterated Expectation

$$\rightarrow E[X] = E[\underbrace{E[X|Y]}_{\text{a RV}}]$$

random version of this

Theorem: Law of Total Variance

$$\rightarrow \text{Var}(X) = E[\text{Var}(X|Y)] + \text{Var} E[X|Y]$$

Ex.  $X|Y=y \sim \text{Bin}(y, p)$ ,  $p \in [0, 1]$   
 $Y \sim \text{Pois}(\lambda)$ ,  $\lambda > 0$

$$E[X]$$

$$(1) E[X|Y=y] = yp$$

$$(2) \underline{E[X|Y] = Yp}$$

$$(3) E[E[X|Y]] = E[Yp] = p EY = \underline{p\lambda}$$

$$\text{Var}(X)?$$

$$\rightarrow \text{Var}(X|Y=y) = yp(1-p)$$

$$\rightarrow \text{Var}(X|Y) = \underline{Yp(1-p)}$$

$$\begin{aligned}\text{Var}(X) &= \text{Var} E[X|Y] + E \text{Var}(X|Y) \\ &= \text{Var}(Yp) + E[Yp(1-p)] \\ &= p^2\lambda + p(1-p)\lambda \\ &= \underline{p\lambda}\end{aligned}$$

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Back to math. stats.

CRLB doesn't always allow us to find UMVUE.

Let's look at how to refine bad estimators into better estimators using conditioning.

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Some facts

① Let  $\hat{\theta}$  be unbiased for  $\tau(\theta)$   
so that  $E\hat{\theta} = \tau(\theta)$

Let  $W = W(\underline{X})$  to be some fun of  $\underline{X}$   
(maybe a stat - maybe not could depend on  $\theta$ )

Let

$$\varphi(w) = \varphi = \mathbb{E}[\hat{\theta} | w]$$

could be a stat - or not

a fn of  $X$

In any case

$$\mathbb{E} \varphi = \mathbb{E}[\mathbb{E}[\hat{\theta} | w]] = \mathbb{E} \hat{\theta} = T(\theta)$$

$$\mathbb{E}[\mathbb{E}[X|Y]] = \mathbb{E} X$$

If  $\varphi$  is a stat, then it is unbiased for  $T(\theta)$ .

$$(2) \text{ Var}(\varphi) \leq \text{Var}(\hat{\theta}).$$

Law of total var says

$$\varphi = \mathbb{E}[\hat{\theta} | w]$$

$$\text{Var} \hat{\theta} = \underbrace{\text{Var} \mathbb{E}[\hat{\theta} | w]}_{\text{Var}(\varphi)} + \underbrace{\mathbb{E} \text{Var}(\hat{\theta} | w)}_{\geq 0}$$

So 
$$\text{Var}(\hat{\theta}) \geq \text{Var}(\varphi).$$

Summary: Start w/  $\hat{\theta}$  where  $\mathbb{E} \hat{\theta} = T(\theta)$

If I define  $\varphi = E[\theta | w]$

then

$$(1) E\varphi = \tau(\theta) \leftarrow$$

$$(2) \text{Var}(\varphi) \leq \text{Var}(\hat{\theta}) \leftarrow$$

Main issue: can't guarantee  $\varphi$  is a stat.

$$\{X_1, X_2\} \stackrel{\text{iid}}{\sim} N(\theta, 1)$$

$$\text{define } \hat{\theta} = \frac{1}{2}(X_1 + X_2)$$

$$\begin{aligned} \text{notice: } E\hat{\theta} &= E\left[\frac{1}{2}(X_1 + X_2)\right] = \frac{1}{2}(EX_1 + EX_2) \\ &= \frac{1}{2}(\theta + \theta) = \theta \end{aligned}$$

$$\text{Var } \hat{\theta} = \frac{1}{4}(\text{Var}(X_1) + \text{Var}(X_2)) = \frac{1}{2}$$

$$w = X_1$$

$$\begin{aligned} \varphi &= E[\hat{\theta} | w] = E\left[\frac{1}{2}(X_1 + X_2) | X_1\right] \\ &= \frac{1}{2}(E[X_1 | X_1] + E[X_2 | X_1]) \end{aligned}$$

$$\text{Aside: } E[z | z=5] = 5$$

1 1 2 1 2 2 7 - 2

$$\downarrow$$
$$= \frac{1}{2}(X_1 + E[X_2])$$

Problem,  $E[X_1 | \theta] = \theta$

$$E[z | z=3] = 3$$

$$\underline{E[z | z] = z}$$

$$= \frac{1}{2} (X_1 + E[X_2])$$

$$= \frac{1}{2} (X_1 + \theta) = \varphi$$

problem:  
not a stat

$$\begin{aligned} \textcircled{1} E[\varphi] &= E\left[\frac{1}{2}(X_1 + \theta)\right] = \frac{1}{2}(E[X_1] + \theta) \\ &= \frac{1}{2}(\theta + \theta) = \theta \end{aligned}$$

$$\textcircled{2} \text{Var}(\varphi) = \frac{1}{4} \text{Var}(X_1) = \frac{1}{4} < \text{Var} \hat{\theta} = \frac{1}{2}$$

Let  $W = \bar{X}$  and re-do

$$\begin{aligned} \varphi &= E[\hat{\theta} | w] = E\left[\frac{1}{2}(X_1 + X_2) | \bar{X}\right] \\ &= \frac{1}{2}(E[X_1 | \bar{X}] + E[X_2 | \bar{X}]) \end{aligned}$$

$$X_n \stackrel{\text{iid}}{\sim} N(\theta, 1) \text{ so } \underline{E[X_1 | \bar{X}] = E[X_2 | \bar{X}] = \dots}$$

$$= \frac{1}{2} 2 E[X_1 | \bar{X}]$$

$$= E[X_1 | \bar{X}]$$

$$= \frac{1}{N} N E[X_1 | \bar{X}]$$

$$= \frac{1}{N} \sum E[X_i | \bar{X}]$$

$$= E\left[\frac{1}{N} \sum_{i=1}^N X_i \mid \bar{X}\right]$$

$$= E[\bar{X} \mid \bar{X}]$$

$$\boxed{\varphi = \bar{X}} \quad (*)$$

this is a stat.

We also know:

$$(1) E\varphi = \theta = E\bar{X}$$

$$(2) \underbrace{\text{Var } \varphi}_{1/N} \leq \underbrace{\text{Var } \hat{\theta}}_{1/2}$$

### Theorem: Rao-Blackwell Theorem

Let  $\hat{\theta}$  is unbiased for  $T(\theta)$  and  $W$  is sufficient for  $\theta$ . Then if

$$\varphi = E[\hat{\theta} \mid W]$$

$$(1) E[\varphi] = T(\theta)$$

$$(2) \text{Var } \varphi \leq \text{Var } \hat{\theta}$$

(3)  $\varphi$  is a statistic.

pf. Fact (3)

... 1. ...

pf. Fact (3)

$$\varphi = E[\underbrace{\hat{\theta}(\underline{x})}_{\bar{\theta}} | W]$$

$$= \int \underbrace{\hat{\theta}(\underline{x})}_{\text{a stat}} \underbrace{f_{\underline{x}|W}(\underline{x})}_{W \text{ sufficient,}}$$

no  $\bar{\theta}$  no  $\bar{\theta}$  in this PDF

$$E[g(\underline{x})] = \int g(\underline{x}) f(\underline{x}) d\underline{x}$$

$$E[g(\underline{x}) | Y=y] = \int g(\underline{x}) f(\underline{x}|y) d\underline{x}$$

Theorem: Lehmann - Scheffé

tech. condition.

Let  $W$  be a (complete) sufficient statistic for  $\theta$  and let  $\hat{\theta}$  be an unbiased est. for  $T(\theta)$  that depends on  $\underline{x}$ s only through  $W$ ,

$$\hat{\theta} = \hat{\theta}(w) = \hat{\theta}(w(\underline{x}))$$

then  $\hat{\theta}$  is the UMVUE for  $T(\theta)$ .

Basically: If I can form an unbiased est. for  $T(\theta)$  from a S.S. it is

the UMVUE.

Ex.  $X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$

Known

don't know

Q: What is the UMVUE for  $\mu$ ?

Use Lehmann - Scheffé

① Find a SS for  $\mu : \bar{X}$

(2) Guess a fn of  $\bar{X}$  that is unbiased for  $\mu$

$\hat{\mu} = \bar{X}$  is unbiased for  $\mu$ ,  $E\bar{X} = \mu$ .

③ So  $\bar{X}$  is the UMVUE.

Ex. let  $T(\mu) = \mu^2$ .

① SS for  $\mu$ :  $\bar{X}$

(2) Find / guess a fn of  $\bar{X}$  that is unbiased for  $\mu^2 = \tau(\mu)$ .



Consider:  $\bar{X}^2$ ,  $E[\bar{X}^2] = \text{Var}(\bar{X}) + (E\bar{X})^2$

$$\boxed{\bar{X} \sim N(\mu, \sigma^2/N)} = \sigma^2/N + \mu^2$$

Let  $\hat{\mu}^2 = \bar{X}^2 - \sigma^2/N$

$$E[\hat{\mu}^2] = E[\bar{X}^2] - \sigma^2/N = \mu^2$$

So  $\bar{X}^2 - \sigma^2/N$  is the UMVUE for  $\mu^2$ .

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Takeaway:

How to find UMVUE for  $T(\theta)$  w/ Lehmann-Scheffé

- ① Find SS for  $\theta$  — call it  $w$
- ② Find a fn of  $w$  that is unbiased for  $T(\theta)$ 
  - (i) Guess fn  $\hat{\theta}(w)$  so that  $E[\hat{\theta}(w)] = T(\theta)$
  - (ii) Use Rao-Blackwell and condition using  $w$ ,  
 $\rightarrow$  Find my unbiased est. of  $T(\theta)$   
call it  $V$   
 $\rightarrow \hat{\theta} = E[V/w]$  is the UMVUE.

