

TodayREVIEW

- One last thought on bias.
- Confidence intervals/tests → when to use which one?
- Normal Approximation to Binomial
→ don't forget the continuity correction!

Example Suppose $\left(\frac{X}{n}\right)^2 = \hat{p}^2$ is used to estimate p^2 , where p = proportion of underlying population in a certain category.

Find the bias in \hat{p}^2 .

Solution We need to find $E(\hat{p}^2) - p^2 = \text{bias}$.

$$E\left(\left(\frac{X}{n}\right)^2\right) = E\left(\frac{\downarrow X^2}{n^2}\right) = \frac{1}{n^2} E(X^2)$$

↑ linear function of r.v. X^2

Cannot use our "linear functions of r.v.s" formulas to find $E(X^2)$.

$$V(X) = E((X-\mu)^2) = E(X^2) - (E(X))^2$$

\uparrow
 $E(X)$

So for any X : $E(X^2) = V(X) + (E(X))^2$

Here $X \sim \text{Bin}(n, p)$

$$\text{So } E(X) = np$$

$$V(X) = np(1-p)$$

\downarrow
 $E(X^2)$

$$= np(1-p) + n^2 p^2.$$

Hence $E\left(\frac{X^2}{n^2}\right) = \frac{1}{n^2} E(X^2) = \frac{1}{n^2} (np(1-p) + n^2 p^2)$

$$\uparrow \qquad \qquad \qquad = \frac{p(1-p)}{n} + p^2$$

$$\text{Bias} = E\left(\left(\frac{X}{n}\right)^2\right) - p^2 = \frac{p(1-p)}{n} + \cancel{p^2} - \cancel{p^2}.$$

Confidence Intervals & Test Statistics

(How to navigate lines 20-29 of Formula Sheet.)

In pairs

C.I.

Test

20) z -C.I. for mean μ 23) z -test for μ

21) t -C.I. for μ 24) t -test for μ

22) (z -) C.I. for proportion p 25) z -test for p

$\left\{ \begin{array}{l} 26) \{ (t-) \text{ C.I. for difference} \\ 27) \text{ of 2 means } \\ \mu_1 - \mu_2 \text{ with variance unknown} \end{array} \right.$
 28) $\{ t$ -test
 29) $\{ \text{for } \mu_1 - \mu_2 \text{ with var. unknown.}$

First check what your C.I. / test is for:

mean μ , proportion p , difference in 2 means $\mu_1 - \mu_2$

① C.I.s/Tests for mean μ

z -tests/C.I.s

(A) Underlying pop: Normal

Variance σ^2 : Known

Sample size n : irrelevant

$$\text{C.I. : } \bar{x} \pm z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}} \quad (20)$$

$$\text{Test : } z_0 = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \quad (23)$$

assumed in H_0

t -tests/C.I.s

(B) Underlying pop: Normal

Variance σ^2 : UNknown

Sample size n : small

$$(n < 40)$$

$$(21) \text{ C.I. : } \bar{x} \pm t_{\frac{\alpha}{2}, n-1} \frac{s}{\sqrt{n}}$$

$$(24) \text{ Test : } t_0 = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$$

(C) Underlying pop: Normal irrelevant
 Variance σ^2 : UNknown (est. because we use CLT.)

Sample size n : large ($n > 40$)

$(s^2$ used to estimate σ^2)

$$\text{C.I. : } \bar{x} \pm z_{\frac{\alpha}{2}} \frac{s}{\sqrt{n}} \quad \xleftarrow{\text{H}_0\text{-value}}$$

$$(\text{Test : } z_0 = (\bar{x} - \mu_0)/s/\sqrt{n})$$

Because of C.L.T., $n > 30$;
 because we est. σ^2 with s^2 , goes up to σ .

Not on formula sheet
 — recover from l. 20
 with s in place of σ

② C.I.s / Tests for proportion

→ No choice ! z - C.I. (22) $\hat{p} \pm z_{\frac{\alpha}{2}} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$

z-test (25) $z_0 = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}} \quad \text{under } H_0$

③ C.I.s / Tests on difference in means $\mu_1 - \mu_2$ v. 0

When variance σ^2 is unknown

→ Always a t-test

Question : Are we allowed to assume $\sigma_1^2 = \sigma_2^2$
 ("variances equal")

or not allowed to assume $\sigma_1^2 = \sigma_2^2$ so
 have to assume $\sigma_1^2 \neq \sigma_2^2$ ("variances unequal")?

Equal : (26.) C.I. : $\bar{x}_1 - \bar{x}_2 \pm t_{\frac{\alpha}{2}, n_1 + n_2 - 2} s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$

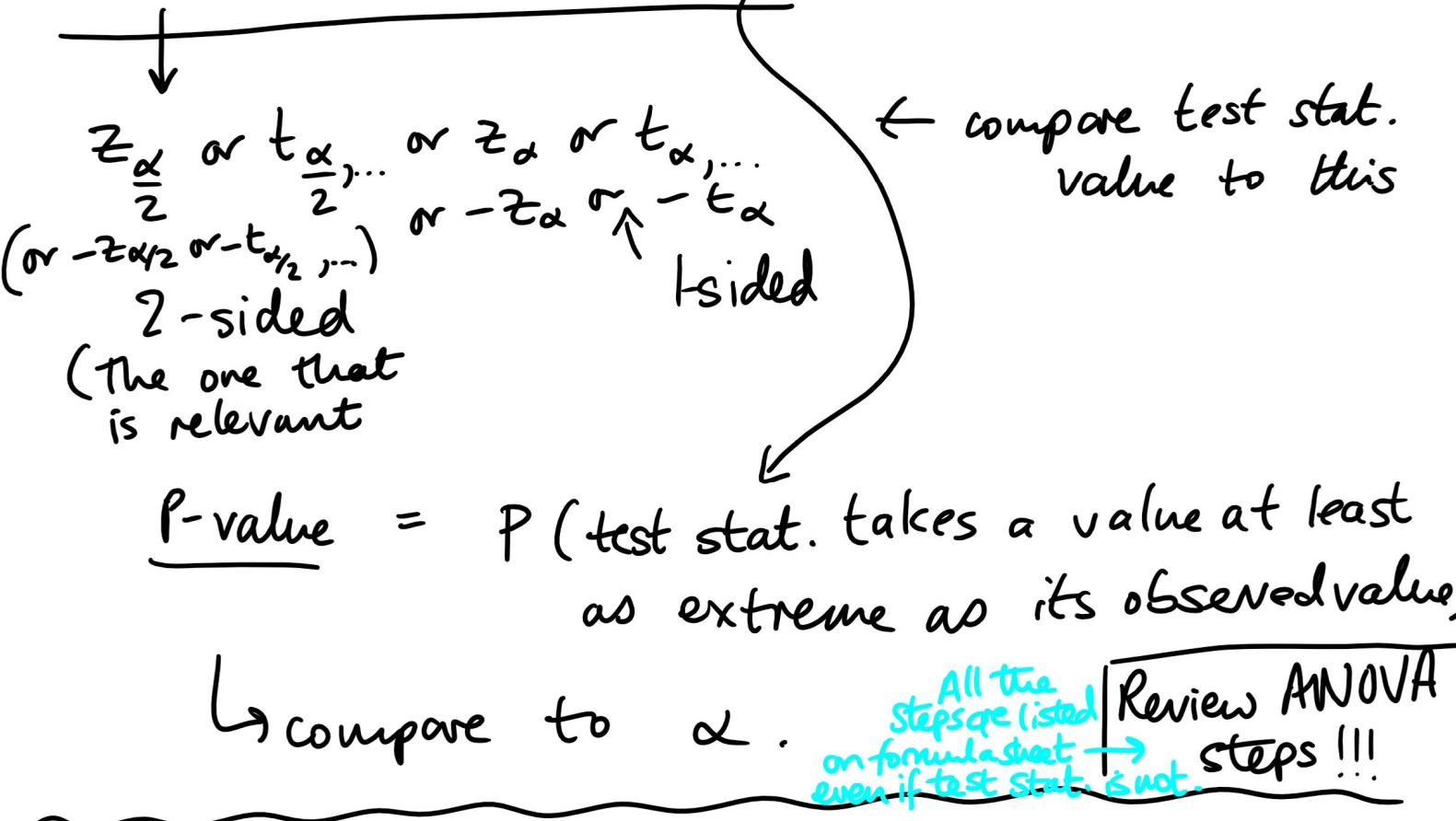
(28.) Test : $t_{n_1 + n_2 - 2} = \frac{\bar{x}_1 - \bar{x}_2}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \quad \begin{matrix} \uparrow \\ s_p^2 = \text{pooled sample variance} \end{matrix}$

Unequal : (27.) C.I. $\bar{x}_1 - \bar{x}_2 \pm t_{\frac{\alpha}{2}, \nu} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$

(28.) Test : $t_\nu = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \quad \begin{matrix} \uparrow \\ \text{degrees of freedom} \end{matrix}$

(27.)

Critical Values & P-Values



Normal Approximation to Binomial

$$X \sim \text{Bin}(n, p)$$

$$\begin{aligned} \text{(mean : } np \\ \text{variance : } np(1-p) \end{aligned}$$

and $np > 5, n(1-p) > 5$ we have :

$$\frac{X - np}{\sqrt{np(1-p)}} \sim N(0, 1).$$

To use this in practice : continuity correction

$$P(X \leq x) = P(X \leq x + 0.5)$$

$$= P\left(Z \leq \frac{(x+0.5)-np}{\sqrt{np(1-p)}}\right)$$

$$P(X=x) = P(X \leq x) - P(X \leq x-1)$$

$$= P\left(Z \leq \frac{(x+0.5)-np}{\sqrt{np(1-p)}}\right) - P\left(Z \leq \frac{(x-0.5)-np}{\sqrt{np(1-p)}}\right)$$

$$P(x_1 \leq X \leq x_2) = P(X \leq x_2) - P(X \leq x_1-1)$$

$$= P\left(Z \leq \frac{x_2+0.5-np}{\sqrt{np(1-p)}}\right) - P\left(Z \leq \frac{x_1-0.5-np}{\sqrt{np(1-p)}}\right)$$

$$= P\left(\frac{(x_1-0.5)-np}{\sqrt{np(1-p)}} \leq Z \leq \frac{x_2+0.5-np}{\sqrt{np(1-p)}}\right)$$

→ Expand the region when making the continuity correction e.g. if the region would be

$[x_1, x_2]$ as in the last example,

expand to $[x_1-0.5, x_2+0.5]$ and then standardize.

e.g. if the region is $\{x\}$, expand to

$[x-0.5, x+0.5]$ & standardize.

& e.g. if the region is $[x, \infty)$ or $(-\infty, x]$

expand to $[x-0.5, \infty)$ or $(-\infty, x+0.5]$.
& standardize.

Final Thoughts:

- ① COURSE EVALUATIONS: I'm sure you're very busy, but please take a 10-minute break before the end of Thursday Dec. 6th to review this course — help us to make it better (as well as tell us what we're doing right)!
- ② You've got this — work hard, but keep your energy levels up for the long haul. Practice, practice, practice — but in the end, be confident you'll do your best, and remember that's anyway all anyone can ask of you. GOOD LUCK!!!