## Ch. 20 - Hypotheses and Test Procedures

Def'n: A <u>null hypothesis</u> is a claim about a population parameter that is assumed to be true until it is declared false.

An <u>alternative hypothesis</u> is a claim about a population parameter that will be true if the null hypothesis is false.

In carrying out a test of  $H_0$  vs.  $H_A$ , the hypothesis  $H_0$  is "rejected" in favour of  $H_A$  only if sample evidence strongly suggests that  $H_0$  is false. If the sample does not contain such evidence,  $H_0$  is "not rejected" or you "fail to reject" it.

NEVER "accept"  $H_0$  or  $H_A$ ...for different reasons.

Ex20.1) 
$$H_0$$
:  $\mu = 2.8$   $H_A$ :  $\mu \neq 2.8$   $\uparrow$  pop'n characteristic hypothesized value or "claim"

Def'n: A two-tailed test has "rejection regions" in both tails.

A <u>one-tailed test</u> has a "rejection region" in one tail.

A <u>lower-tailed test</u> has the "rejection region" in the left tail. An upper-tailed test has the "rejection region" in the right tail.

Ex20.2)

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a)	$H_0$ : $\mu = 15$	$H_A$ : $\mu = 15$	→ INCORRECT
b)	$H_0$ : $\mu = 123$	$H_A$ : $\mu = 125$	→ INCORRECT
c)	$H_0$ : $\mu = 123$	$H_A$ : $\mu < 123$	→ CORRECT
d)	H <sub>0</sub> : $\mu \ge 123$	$H_A$ : $\mu < 123$	→ CORRECT
e)	$H_0$ : $p = 0.4$	$H_A$ : $p > 0.6$	→ INCORRECT
f)	$H_0$ : $p = 1.5$	$H_A$ : $p > 1.5$	→ INCORRECT
g)	$H_0$ : $\hat{p} = 0.1$	$H_A$ : $\hat{p} \neq 0.1$	→ INCORRECT

	Two-Tailed Test	<b>Lower-Tailed Test</b>	<b>Upper-Tailed Test</b>
Sign for H <sub>0</sub>	=	= or ≥	= or ≤
Sign for H <sub>A</sub>	<i>≠</i>	<	>
"Rejection region"	In both tails	In the left tail	In the right tail

Ex20.3)

Is the mean different than $\mu_0$ ?	$H_0$ : $\mu = \mu_0$	$H_A$ : $\mu \neq \mu_0$
Is the mean lower than $\mu_0$ ?	H <sub>0</sub> : $\mu \ge \mu_0$	$H_A$ : $\mu < \mu_0$
Is the mean lower or still the same than $\mu_0$ ?	H <sub>0</sub> : $\mu \le \mu_0$	$H_A$ : $\mu > \mu_0$
Is the mean higher than $\mu_0$ ?	H <sub>0</sub> : $\mu \le \mu_0$	$H_A$ : $\mu > \mu_0$

Def'n: A <u>test statistic</u> is the function of the sample data on which a conclusion to reject or fail to reject  $H_0$  is based. For example, Z and t are test statistics.

The <u>P-value</u> is a measure of inconsistency between the hypothesized value for a pop'n characteristic and the observed sample. Assuming  $H_0$  is true, the <u>P-value</u> can be defined as the probability of obtaining a test statistic value at least as inconsistent with  $H_0$  as what actually resulted. Keep in mind that we *want* to be inconsistent with  $H_0$  to reject it. **Thus, the smaller the P-value, the more likely we reject H\_0**.

The <u>significance level</u> (denoted by  $\alpha$ ) is a number such that we reject H<sub>0</sub> if the *P*-value is less than or equal to that number.

The "significance level approach":

reject H<sub>0</sub> if 
$$p$$
-value  $\leq \alpha$  do not reject H<sub>0</sub> if  $p$ -value  $> \alpha$ 

Common choices for  $\alpha$  are 0.01, 0.05, and 0.1, depending on the nature of the test.

#### PROBLEMS:

- a) If you're comparing to  $\alpha = 0.05$ , are the *P*-values 0.045 and 0.000 001 "different"?
- b) If we use a "cut-off" like  $\alpha = 0.05$ , does it make sense to conclude differently between *P*-values of 0.049 and 0.051?

Solution: ALWAYS report your *P*-value! That way a reader may draw their own conclusions. Moreover, use the "judgment approach" for rejection. Here, there's a tendency of avoiding "cut-off" points and going toward some "acceptable" guidelines:

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0.01 > P-value > 0 \rightarrow strong to convincing evidence against H<sub>0</sub>
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0.05 > P-value  $> 0.01 \rightarrow$  moderate to strong evidence against H<sub>0</sub>

0.10 > P-value  $> 0.05 \rightarrow$  suggestive to moderate evidence against H<sub>0</sub>, yet inconclusive

1 > P-value > 0.1  $\rightarrow$  weak evidence against  $H_0$ 

### Steps of a Significance Test:

- 1. Assumptions: Specify variable/parameter. What assumptions apply? Do they hold?
- 2. *Hypotheses*: State the null/alternative hypotheses. (Select  $\alpha$  for the test.)
- 3. *Test statistic*: Use the appropriate formula for the given situation.
- 4. *P-value*: Determine an exact value or range.
- 5. *Conclusion*: Make a decision and conclude within the context of the problem.

# Significance Tests About Proportions

Recall the 3 rules from Chapter 18. They collectively imply that when n is large,

$$z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}} \text{ is approximately } N(0, 1).$$

Assumptions: categorical variable, random sample,  $np_0 \ge 15$  and  $n(1 - p_0) \ge 15$ .

Ex20.4) Recall the survey of random people from Ex19.3). Suppose p is 0.240. Does the sample disprove this claim? Test the claim using both approaches and  $\alpha = 0.01$ .

$$np_0 = 1356(0.240) = 325.44$$
  $n(1-p_0) = 1356(1-0.240) = 1030.56$ 

Since both values are greater than 15, the sample size is large. Consequently, we may use a normal distribution. (Note also the sample is random.)

$$H_0$$
:  $p = 0.240$   $H_A$ :  $p \neq 0.240$ 

$$z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}} = \frac{0.265 - 0.240}{\sqrt{\frac{0.240(1 - 0.240)}{1356}}} = \frac{0.025}{0.0116} = 2.13$$

$$P$$
-value =  $2*P(Z > 2.13) = 2*(1 - 0.9834) = 2*(0.0166) = 0.0332$ 

Using the "significance level approach", we have  $\alpha = 0.01 < 0.0332$ , so we do not reject  $H_0$ . Using the "judgment approach", we have moderate to strong evidence against  $H_0$ . Using SLA, the claimed population proportion could still be valid, but the JA concludes that the claimed value is inappropriate.

(Note that one-tailed tests with this example were also discussed in class.)

#### **Summary**

Each hypothesis test should include:

- clear null & alternative hypotheses
- assumptions (stated and checked)
- appropriately-used test statistic (show the formula, identify its distribution)
- calculation of both the test statistic and *P*-value (exact or range)
- conclusion in the context of the problem

## Ch. 21 - Errors in Hypothesis Testing

In any hypothesis test, there is 1 of 2 choices: reject or not reject. There is also 1 of 2

choices as the test applies to reality:  $H_0$  is true or  $H_0$  is false.

		Actual situation	
		H <sub>0</sub> is true	H <sub>0</sub> is false
	Do not reject H <sub>0</sub>	Correct	Type II or
Decision		Decision	$\beta$ error
Decision	Reject H <sub>0</sub>	Type I or	Correct
		$\alpha$ error	Decision

Def'n: A Type I error occurs when a true null hypothesis is rejected. The value of  $\alpha$ represents the prob. of committing this type of error; that is,

$$\alpha = P(H_0 \text{ is rejected} \mid H_0 \text{ is true})$$

The value of  $\alpha$  represents the *significance level* of the test.

A Type II error occurs when a false null hypothesis is not rejected. The value of  $\beta$ represents the prob. of committing a Type II error; that is,

$$\beta = P(H_0 \text{ is not rejected} \mid H_0 \text{ is false})$$

The value of  $1 - \beta$  is called the *power of the test*. It represents the probability of NOT making a Type II error. Or, power =  $P(\text{rejecting H}_0 \mid \text{H}_0 \text{ is false})$ .

Ex21.2) H<sub>0</sub>: "innocent until proven guilty"

, .		Actual situation	
		Innocent	Guilty
	Find not guilty	Correct	Type II or
Inmy's desision		Decision	$\beta$ error
Jury's decision	Find guilty	Type I or	Correct
		$\alpha$ error	Decision

These 2 errors are dependent. For a fixed sample size, lowering  $\alpha$  will raise  $\beta$  and vice versa. Decreasing  $\alpha$  and  $\beta$  simultaneously requires increasing the sample size. Further information on the errors are not covered in this course.