

## 10.7 Comments on Hypothesis Testing

Note: When the test statistic does not fall in the RR, we do not say that we \_\_\_\_\_  $H_0$ .

- Rather, we \_\_\_\_\_  $H_0$ .
- When we \_\_\_\_\_, say,  $H_0: \theta = \theta_0$ , it does not imply that we are concluding that the true value of  $\theta$  is  $\theta_0$ .
- In fact,  $\theta$  might not equal  $\theta_0$ , but our test does not detect this difference (possibly because of a \_\_\_\_\_ or because the difference between  $\theta$  and  $\theta_0$  is \_\_\_\_\_).

Note: When a hypothesis test rejects  $H_0$ , we say the result is \_\_\_\_\_.  
- We should not confuse \_\_\_\_\_ with \_\_\_\_\_.

- Suppose we test  $H_0: \mu_1 = \mu_2$  vs.  $H_a: \mu_1 \neq \mu_2$ , comparing mean reaction times between two populations, and we have very large samples.
- We might get a p-value of 0.02, implying \_\_\_\_\_ significance if  $\alpha = 0.05$ .
- But if the sample mean reaction times only differ by 0.2 seconds, this difference may not be \_\_\_\_\_ significant.
- A CI for  $\mu_1 - \mu_2$  can better assess the practical size of the difference in this case.

### 10.8 Small-sample Tests for $\mu$ and for $\mu_1 - \mu_2$

- Section 10.3 covered tests for  $\mu$  and  $\mu_1 - \mu_2$  when the sample size(s) were large.
- Now suppose we have  $Y_1, Y_2, \dots, Y_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$  where  $\mu$  and  $\sigma^2$  are unknown and  $n$  may be small.
- Consider testing  $H_0: \mu = \mu_0$  against one of the three typical alternatives.

- If  $H_0: \mu = \mu_0$  is true, then the statistic has a \_\_\_\_\_ with \_\_\_\_\_ (shown in STAT 512).

- The test procedures in this situation are very similar to the z-tests from Sec. 10.3, but now we use:

<u>Small-sample Test about <math>\mu</math></u>			
<u><math>H_0</math></u>	<u><math>H_a</math></u>	<u>Test Statistic</u>	<u>RR</u>

- Example 1: If the mean breaking strength for a manufactured ribbon is less than 185 pounds, it is too weak to sell.

An inspector selects 12 ribbons and tests their breaking strengths, finding  $\bar{Y} = 180.4$  and  $s = 8.21$  for this sample.

Test

using  $\alpha = .05$ .

### Small-sample Test about $\mu_1 - \mu_2$

- Recall that if we have two independent samples  $Y_{11}, Y_{12}, \dots, Y_{1n_1}$   $\stackrel{iid}{\sim} N(\mu_1, \sigma^2)$  and  $Y_{12}, Y_{22}, \dots, Y_{2n_2}$   $\stackrel{iid}{\sim} N(\mu_2, \sigma^2)$ , and if

then

- We can then test  $H_0: \mu_1 - \mu_2 = \Delta_0$  with a two-sample t-test:

Example 2: A professor taught her two calculus classes using two different teaching methods (one traditional, one modern). She wanted to test whether the two methods produced significantly different mean final exam scores. The "traditional" class yielded  $\bar{Y}_1 = 78.63$  and  $s_1^2 = 24.62$  (with  $n_1 = 12$ ) and the "modern" class ( $n_2 = 18$ ) yielded  $\bar{Y}_2 = 80.95$ ,  $s_2^2 = 23.91$ . Test using  $\alpha = .05$ .

- The t-tests are robust: If the assumptions behind them are moderately wrong, they still work approximately well:
- If the sample(s) are moderately non-normal, the t-tests can be used safely.
- If  $\sigma_1 \neq \sigma_2$ , the two-sample t-test can be used safely when  $n_1 \approx n_2$ .

## 10.9 Tests About Variances

- We now consider tests related to population variances.
- Let  $Y_1, Y_2, \dots, Y_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$  with  $\mu$  and  $\sigma^2$  unknown.
- We test  $H_0: \sigma^2 = \sigma_0^2$  against either an upper-tail, lower-tail, or two-tailed alternative.
- Recall that if  $H_0: \sigma^2 = \sigma_0^2$  is true, then
  - Let us use \_\_\_\_\_ as our test statistic.
  - Suppose we have  $H_a: \sigma^2 > \sigma_0^2$ .
  - Then \_\_\_\_\_ values of  $S^2$  favor  $H_a$ .  
⇒ Reject  $H_0$  when  $S^2$  is \_\_\_\_\_.  
⇒ Reject  $H_0$  when

- Then the rejection region that keeps  
 $P[\text{Reject } H_0 \mid H_0 \text{ true}] = \alpha$  is:

Picture:

Similarly, for  $H_a: \sigma^2 < \sigma_0^2$ ,

For  $H_a: \sigma^2 \neq \sigma_0^2$ ,

Picture:

Example 1: A random sample of 20 sheet metal width measurements yielded a sample variance of 0.64. (Assume the data are normally distributed.) Test  $H_0: \sigma^2 = 0.6$  vs.  $H_a: \sigma^2 \neq 0.6$  using  $\alpha = 0.05$ .

### Test to Compare Two Variances

- If we have two independent samples (of sizes  $n_1$  and  $n_2$ ) from  $N(\mu_1, \sigma_1^2)$  and  $N(\mu_2, \sigma_2^2)$  populations, we can formally compare  $\sigma_1^2$  and  $\sigma_2^2$ .
- Recall that
- So dividing by the respective d.f. and taking the ratio, we see:

has an \_\_\_\_\_ with \_\_\_\_\_  
and \_\_\_\_\_.

- If  $H_0: \sigma_1^2 = \sigma_2^2$  is true, then

- So

can be our test statistic.

- If  $H_a: \sigma_1^2 > \sigma_2^2$ , then \_\_\_\_\_ values of F favor  $H_a$ .

- RR:

- If  $H_a$  is in the reverse direction, simply reverse the labeling of populations 1 and 2 and conduct the upper-tail test.

- If  $H_a: \sigma_1^2 \neq \sigma_2^2$ , then \_\_\_\_\_ values of F favor  $H_a$ .

- Note that both F and  $\frac{1}{F}$  have F-distributions, but with the d.f. reversed.

- So we can reject  $H_0$  if

RR:

Example 2: Suppose two (independent) normal samples of 11 and 13 measurements yielded  $s_1 = 8.62$  and  $s_2 = 9.13$ . Test  $H_0: \sigma_1^2 = \sigma_2^2$  vs.  $H_a: \sigma_1^2 \neq \sigma_2^2$  using  $\alpha = .05$ .