

## 10.10 Power and the Neyman-Pearson Lemma

- Recall that  $\beta = P[\text{Type II error}]$

=

- For a particular value of a parameter that is in the alternative region (say  $\theta_a$ ), we can calculate  $\beta$  at  $\theta_a$  and we can define the power of the test at  $\theta_a$  as

- In general, the power function of a test is a function of  $\theta$  and can be defined for any value of  $\theta$  as

- If we test  $H_0: \theta = \theta_0$ , then the power function evaluated at  $\theta_0$  is

- But for any  $\theta_a$  in the alternative region, then  $\text{Power}(\theta_a)$  is

which we would like to be \_\_\_\_\_.

- Since the power at  $\theta_a$  is  $1 - \beta$  at  $\theta_a$ , we can calculate the power of a test similarly as we calculated  $\beta$  back in Section 10.4.

- To calculate the power function, we would have to calculate the power at all values of  $\theta$  (tedious!) so we can use a computer to help.

Recall Example 1, Sec. 10.4 notes: We tested

$H_0: \mu = 500$  vs.  $H_a: \mu > 500$  (at  $\alpha = .05$ ) with a  $z$ -test. (Assume  $\sigma \approx 97.4$ ).

- We earlier calculated  $\beta = .6064$  when  $\mu = 515$ .  
So  $\text{Power}(515) =$

## Picture of Power Function:

- Note that for parameter values in the null region ( $\mu \leq 500$ ), the maximum  $P[\text{Reject } H_0]$  is achieved at the \_\_\_\_\_ of the null region (at \_\_\_\_\_).
- This is generally true for the tests we consider, and it motivates writing  $H_0: \mu = 500$  instead of  $H_0: \mu \leq 500$ .
- Same example, but suppose we test  $H_0: \mu = 500$  vs.  $H_a: \mu \neq 500$  at  $\alpha = .05$ .

## Picture of Power Function:

- The common tests we study have power functions similar to these examples.
- We see the farther the true parameter value is away from the null value (or region), the \_\_\_\_\_ power the test has to reject  $H_0$ .
- An ideal test would have
  - (1) Power  $\leq \alpha$  for parameter values in the \_\_\_\_\_ region
  - (2) Power as high as possible for parameter values in the \_\_\_\_\_ region.
- Such a test is called a most-powerful  $\alpha$ -level test.

# Simple and Composite Hypotheses

Defn: A statistical hypothesis is called simple if it completely specifies the distribution from which the sample is taken

Defn: A hypothesis that does not completely specify the distribution of the data is called composite.

Example 1: Suppose  $Y_1, \dots, Y_n \stackrel{iid}{\sim} \text{Expon}(\beta)$ . Then:

$H: \beta = 2$  is a \_\_\_\_\_ hypothesis.

$H: \beta > 2$  is a \_\_\_\_\_ hypothesis.

$H: \beta \geq 2$  is a \_\_\_\_\_ hypothesis.

Example 2(a): Suppose  $Y_1, \dots, Y_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$  with  $\mu, \sigma^2$  unknown. Then

$H: \mu = 3$  is a \_\_\_\_\_ hypothesis.

Example 2(b): Suppose  $Y_1, \dots, Y_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$  with  $\sigma^2$  known to be 1. Then:

$H: \mu = 3$  is a \_\_\_\_\_ hypothesis.

$H: \mu \neq 3$  is a \_\_\_\_\_ hypothesis.

## The Neyman-Pearson Lemma

- Suppose we are testing a simple null hypothesis versus a simple alternative hypothesis:

$$H_0: \theta = \theta_0 \text{ vs. } H_a: \theta = \theta_a$$

where  $\theta_0$  and  $\theta_a$  are two numbers and our data  $Y_1, \dots, Y_n$  are iid from a distribution having parameter  $\theta$ .

Neyman-Pearson Lemma: For a given  $\alpha$ , the test with the highest power at  $\theta_a$  has a rejection region of the form

where  $k$  is a constant chosen to maintain

- The N-P Lemma allows us to find the most-powerful (MP)  $\alpha$ -level test for a simple-vs.-simple situation.

Proof (continuous case): Let  $C$  be the "rejection region" corresponding to the Neyman-Pearson test; i.e., let  $C$  be the set of sample points that lead us to reject  $H_0$ . Then the power of the N-P test at  $\theta_a$  is:

- Let  $D$  be the "rejection region" of any other  $\alpha$ -level test.

Note:

- The proof in the discrete case is very similar.