

Chapter 9-10

Confidence Intervals and Hypothesis Testing

Type 2 Error, Power

Statistics

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Type 1 Error (α)

- Any other way to decrease type 1 error? (*other than increasing n*)
- We may change the limits of the critical regions.
- For example change them to
 - $\bar{X} > 73$
 - $\bar{X} < 63$
- . Is that OK?
 - Then, you make almost no error if you reject H_0
 - Calculate!
 - $\alpha = P\left(\frac{\bar{X}-\mu}{0.45} < \frac{63-68}{0.45} \mid \mu = 68\right) + P\left(\frac{\bar{X}-\mu}{0.45} > \frac{73-68}{0.45} \mid \mu = 68\right) = 0$ Any problem?

Type 2 Error (β)

- However, what happens if you **don't reject** H_0
- Then you purchase the cases whose
 - μ is 66 mm
 - or whose μ is 70 mm.
- This is something you don't want!!!
- Let's illustrate this

Type 2 Error (β)

- Consider the case where you decided that $\mu = 68mm$ even if the real parameter is $\mu = 70mm$
- Hence you decided to purchase a large amount of these cases.
- Find the probability of this situation, i.e.
 - $P(\text{buy a large amount of cases when in fact } \mu = 70 \text{ mm})$
 - $= P(\text{Do not reject } H_0 \mid \mu = 70)$
 - $= P(63 \leq \bar{X} \leq 73 \mid \mu = 70 \text{ mm})$

Type 2 Error (β)

- $= P(63 \leq \bar{X} \leq 73 \mid \mu = 70)$
- $= P\left(\frac{63-\mu}{0.45} \leq \frac{\bar{X}-\mu}{0.45} \leq \frac{73-\mu}{0.45} \mid \mu = 70\right)$
- $= P\left(\frac{63-\mu}{0.45} \leq \frac{\bar{X}-\mu}{0.45} \leq \frac{73-\mu}{0.45} \mid \mu = 70\right)$
- $= P\left(\frac{63-70}{0.45} \leq \frac{\bar{X}-70}{0.45} \leq \frac{73-70}{0.45} \mid \mu = 70\right)$
- $= P(-15.55 \leq Z \leq 6.66) = 1$
- This probability is 1!!!

Type 2 Error (β)

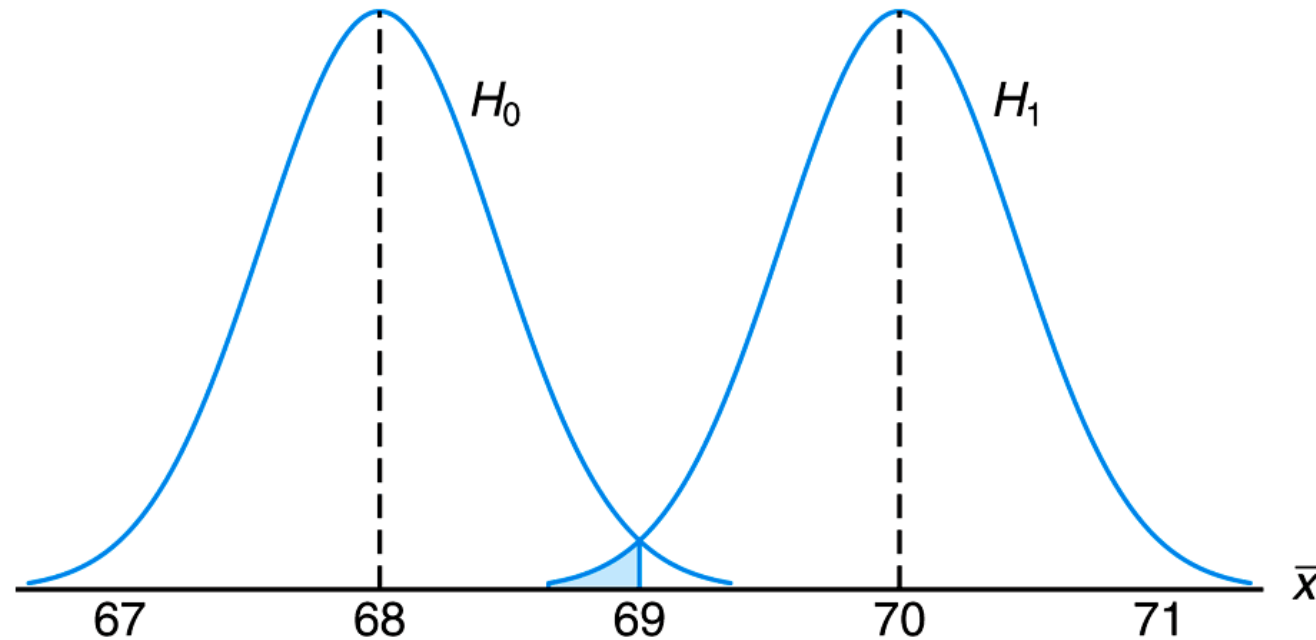
- Hence, if you change your rejection region from
 - $\bar{X} > 69$ or $\bar{X} < 67$
 - to
 - $\bar{X} > 73$ or $\bar{X} < 63$
- Then, if the real $\mu = 70 \text{ mm}$ and if you don't reject the hypothesis, then you are **almost always** making an error!
- This error is indeed **type 2 error**:
 - *Do not reject H_0 when it is not true.*
- Recall that $H_0: \mu = 68$ and we calculated the following:
 - $\beta = P(\text{Do not reject } H_0 \mid \mu = 70)$

Type 2 Error (β)

- In order to calculate β , we need to specify a particular value for μ
 - “ H_0 is not true” is not sufficient to calculate a probability.
 - In the previous example we calculate it for $\mu = 70$
- Hence we define β as $\beta(\mu)$ in order to indicate that it depends on the value of selected μ .
- Set the critical region to original one ($\bar{X} > 69$ or $\bar{X} < 67$) and calculate β for the following:
 - the **true mean is $\mu = 70$**
 - the **true mean is $\mu = 66$**
 - the **true mean is $\mu = 68.5$**

Probability of a Type II Error

Let's now calculate β , the probability of committing a type II error, assuming that the true value of $\mu = 70$, for the case $n=64$.

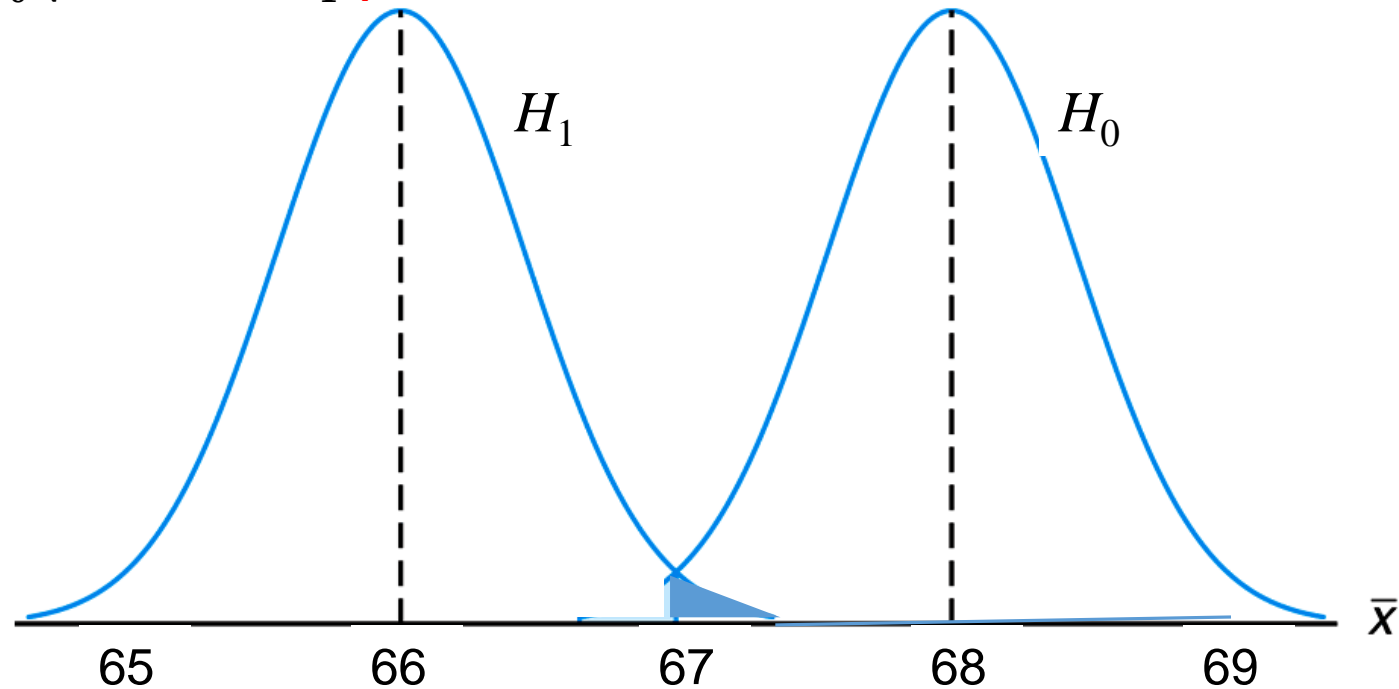


$$\begin{aligned}\beta &= P(\text{type II error} | \mu = 70) = P(67 \leq \bar{X} \leq 69 | \mu = 70) \\ &= P\left(\frac{67 - 70}{0.45} \leq Z \leq \frac{69 - 70}{0.45}\right) \approx P(-6.67 \leq Z \leq -2.22) \approx 0.0132.\end{aligned}$$

Probability of a Type II Error

Now calculate it for $\mu = 66$, that is:

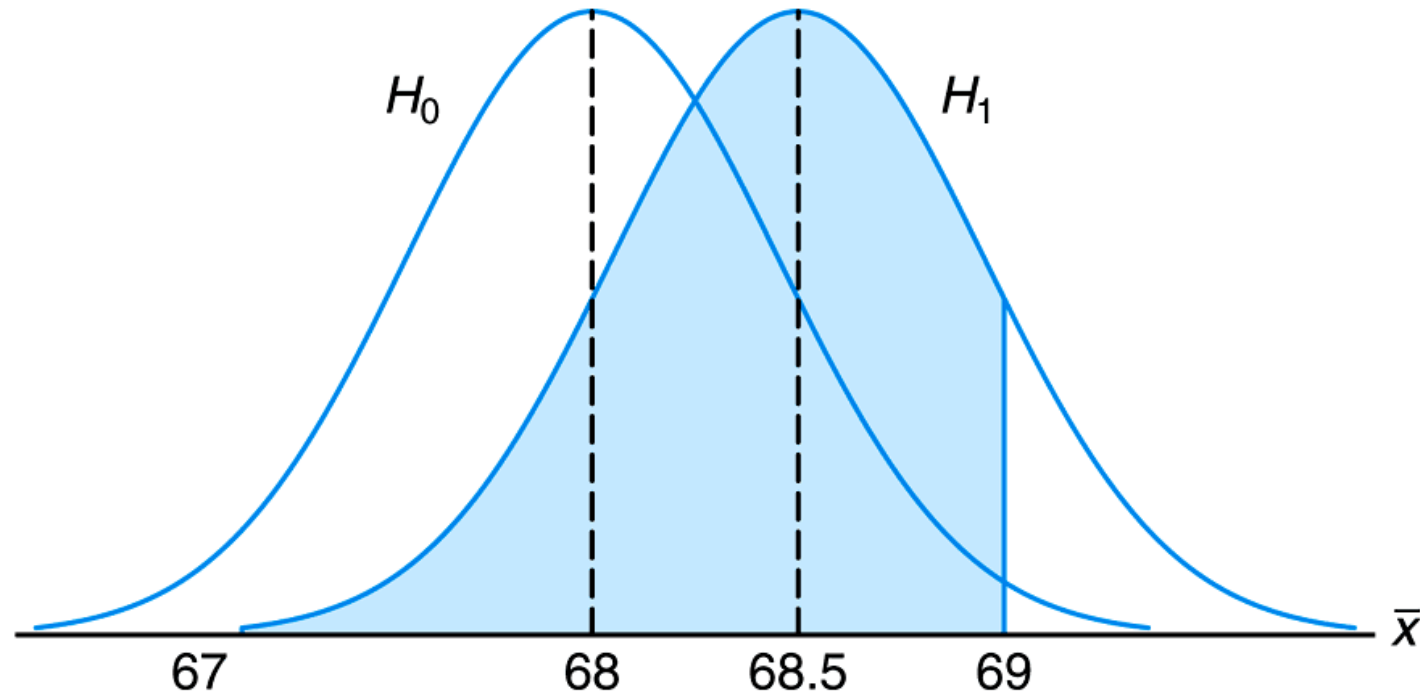
$H_0: \mu = 68$ vs $H_1: \mu = 66$ with $n = 64$.



$$\begin{aligned}
 b(70) &= P(\text{type II error} \mid m = 66) = P(67 \leq \bar{X} \leq 69 \mid m = 66) \\
 &= P\left(\frac{67 - 66}{0.45} \leq Z \leq \frac{69 - 66}{0.45}\right) \gg P(2.22 \leq Z \leq 6.67) \gg 0.0132.
 \end{aligned}$$

Probability of a Type II Error

$$H_0 : m = 68 \quad \text{vs.} \quad H_1 : m = 68.5$$



$$\begin{aligned} b(68.5) &= P(\text{type II error} \mid m = 68.5) = P(67 \leq \bar{X} \leq 69 \mid m = 68.5) \\ &= P\left(\frac{67 - 68.5}{0.45} \leq Z \leq \frac{69 - 68.5}{0.45}\right) \gg P(-3.33 \leq Z \leq 1.11) \gg 0.866. \end{aligned}$$

Type 2 Error (β)

- We can generalize this to see that β values will be the same when μ is equal distance from 68 ($H_0: \mu = 68$).
- We also note that β
 - becomes smaller as $|\mu - 68|$ increases, i.e., μ gets away from 68.
 - becomes larger as $|\mu - 68|$ decreases, i.e., μ gets closer to 68.
- This is understandable as it becomes more and more difficult to **identify the difference** between two means when $|\mu - 68|$ nears zero, causing the probability of a type II error to increase.

Type 1 and Type 2 Errors

- For fixed sample size, when one error type increases, the other error type decreases.
- An increase in the sample size n will reduce both α and β .

Testing a Statistical Hypothesis: Errors

- Good to control the probability of these two types of errors.
- Unfortunately, for a given sample size, we can NOT control both
- We use a hypothesis testing procedure that controls the probability of a type I error and guarantees that it cannot exceed a given small probability α .

Testing a Statistical Hypothesis: Errors

- **Example (Cont.d):** For the previous example, instead of the critical region $\bar{X} < 67$ and $\bar{X} > 69$, what is probability of type 1 error if we are given $\alpha = 0.05$.
- **Solution**
- Let the critical region be $\bar{X} < c_1$ and $\bar{X} > c_2$
- First find the critical region given $\alpha = 0.05$, i.e.,
 - Find c_1 and c_2 such that, given H_0 is true, the following must hold:
 - $P(\bar{X} < c_1) + P(\bar{X} > c_2) = 0.05$

Testing a Statistical Hypothesis: Errors

- Since they are symmetric we have
 - $P(\bar{X} > c_2) = 0.025$
- Using CLT and $\sigma_{\bar{X}} = 0.45$ we have
 - $c_2 = \mu + z_{0.025} \times 0.45 = 68.74$
- Similarly we can find $c_1 = 67.26$
- Now calculate $P(\text{type 1 error})$
 - $= P(\text{reject } H_0 | H_0 \text{ is true})$
 - $= P(\bar{X} > 68.74 \text{ or } \bar{X} < 67.26 | \mu = 68)$
 - $= P(\bar{X} > 68.74 | \mu = 68) + P(\bar{X} < 67.26 | \mu = 68)$
 - $= 0.025 + 0.025 = 0.05$

Testing a Statistical Hypothesis: Errors

- *Well, in fact α value that we used since the beginning is nothing but type 1 error.*
- Now calculate β when the real $\mu = 70$
- $\beta(70) = P(67.26 < \bar{X} < 68.74 | \mu = 70)$
- Using CLT and $\sigma = 0.45$ we have
- $= P\left(\frac{67.26 - \mu}{0.45} \leq \frac{\bar{X} - \mu}{0.45} \leq \frac{68.74 - \mu}{0.45} \mid \mu = 70\right)$
- $= P\left(\frac{67.26 - 70}{0.45} \leq \frac{\bar{X} - 70}{0.45} \leq \frac{68.74 - 70}{0.45}\right) = 0.0026$
- Hence probability of type 2 error is very small for this example.

Power of a Test

- **DEFINITION.** $1-\beta$ = **Power** of the test
 $= P(\text{Reject } H_0 \text{ when } H_0 \text{ is false})$
- Power of the test = The probability of **correctly** rejecting a false H_0 .
- It is the power of the test to differentiate between two hypotheses
 - If two hypotheses are close, then the power will be low.

Power of a Test

- $1 - \beta$ = **Power** of the test = $P(\text{Reject } H_0 \text{ when } H_0 \text{ is false})$
- A good testing procedure should have
 1. A **small level of significance** (α value), and
 2. A **large power** ($1 - \beta$ value) when the deviation from the value specified by H_0 is significantly large.
- For the previous example, the power of the test is $1 - \beta = 0.9974$

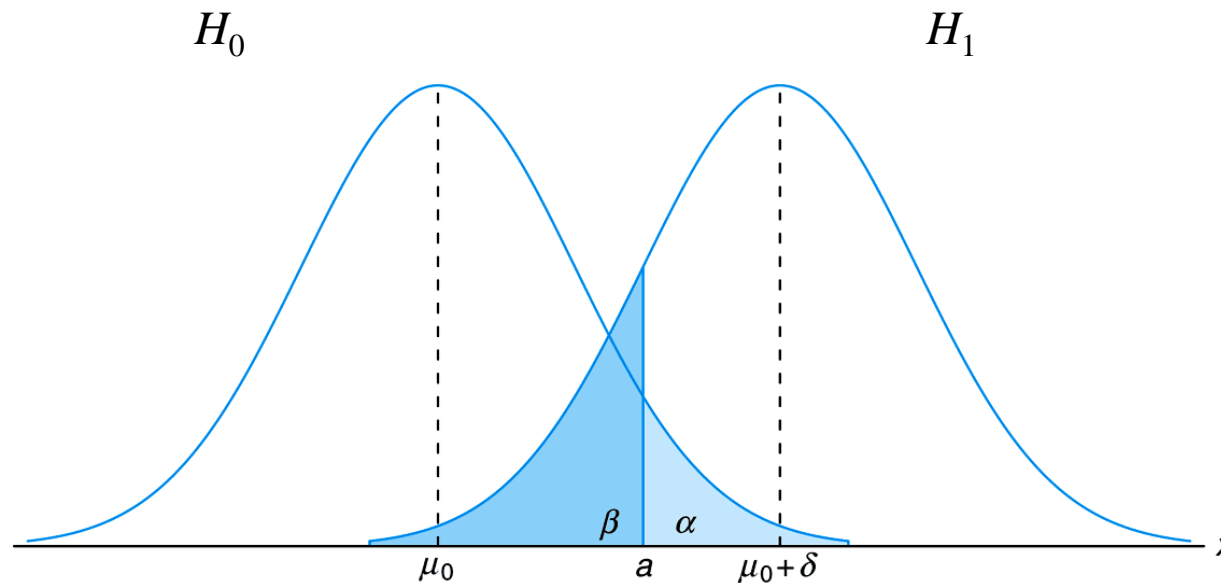
Power of a Test

- $1-\beta$ = **Power** of the test = $P(\text{Reject } H_0 \text{ when } H_0 \text{ is false})$
- You should now practice the calculation of the power of this test for other values of μ under H_1 .
- In particular calculate $1-\beta(67.5)$ to see how it is done in EXCEL.
- **EXERCISE.** Prepare a simple EXCEL sheet for calculating the power of the test in Example 2 for many different values of μ under H_1 . Then graph the power function $1-\beta(\mu)$ against μ values.

Testing $\mu = \mu_0$ versus $\mu = \mu_0 + \delta$

- When testing $H_0: \mu = \mu_0$ against $H_1: \mu = \mu_0 + \delta$, for $\delta > 0$ when σ is given, we can use the Z-test.
- For given α, β, δ the needed sample size is:
$$n \gg \frac{(z_a + z_b)^2 S^2}{d^2}$$

Sometimes we need the number of experiments to be made for a given value of type 1 and type 2 error values. For the **single** mean case, it is given by the above formula.



Testing $\mu_1 - \mu_2 = d_0$ versus $\mu_1 - \mu_2 = d_0 + \delta$

- When testing $H_0: \mu_1 - \mu_2 = d_0$ against $H_1: \mu_1 - \mu_2 = d_0 + \delta$, when σ is given.
- we can use the Z-test, taking the sample size as:
$$n \gg \frac{(z_a + z_b)^2 (S_1^2 + S_2^2)}{d^2}$$

Sometimes we need the number of experiments to be made for a given value of type 1 and type 2 error values. For the **two population** case, it is given by the above formula.

