

## Lecture 21 (Ch. 8)

Last time: **Belief**: a statement NOT based on data ("a priori")

Example Data says:  $n=64$ ,  $\bar{x}_{obs}=34.4$ ,  $S=1.1$ .

Does the data provide evidence to support  $\mu_x \geq 34$ ?

→ So, the a priori belief is  $\mu_x < 34$ .

Let's assume the belief is true

Formally, Let  $H_0: \mu_x < 34$ . Assume  $H_0 = \text{True}$ .

Evidence, from data, against (Contrary to) This assumption

$$p\text{-value} = \text{pr}(\bar{x} > \bar{x}_{obs} \mid \mu_x = 34)$$

↑  
against (Contrary to)  $H_0$

↑ sufficient to test equality.  
No need to do  $\mu_x < 34$   
(See the FYI fig, below).

$$= \text{pr}\left(\frac{\bar{x} - \mu_x}{S/\sqrt{n}} > \frac{\bar{x}_{obs} - \mu_x}{S/\sqrt{n}}\right) = \text{pr}(t > t_{obs}) \approx .0025$$

$$\equiv t_{obs} = \frac{34.4 - 34}{1.1/\sqrt{64}} = 2.71$$

This p-value is small!

It's really easy to misinterpret This small prob.

But the best way to think of it is that it suggests our assumption was bad, and so, we should reject it.

In short: **small p-value**  $\Rightarrow$  large evidence against  $H_0$ .

↓  
**Reject  $H_0$**

What determines "small"?

$\alpha$  does = 1 - conf. level.

And you choose  $\alpha$ .

Here is the generalization of the above example:

Dropping subscript  $x$ .

Example

$\mu_x, \pi_x, \dots$

1) Decide the pop. parameter being tested

2) Set-up  $H_0$  (Null hyp.) and  $H_1$  (Alternative hyp.)

$$\begin{cases} H_0: & \mu > \mu_0 \\ H_1: & \mu < \mu_0 \end{cases}$$

$$\mu < \mu_0$$

$$\mu = \mu_0$$

$$\mu < 34 = \mu_0$$

$$\mu > 34 \neq \mu_0$$

$$\begin{cases} H_0: & \mu < \mu_0 \\ H_1: & \mu > \mu_0 \end{cases}$$

$$\mu > \mu_0$$

$$\mu \neq \mu_0$$

$$\begin{cases} H_0: & \mu = \mu_0 \\ H_1: & \mu < \mu_0 \end{cases}$$

$$\mu = \mu_0$$

$$\mu = \mu_0$$

sufficient to test equality in  $H_0$ .

$$\begin{cases} H_0: & \mu < \mu_0 \\ H_1: & \mu > \mu_0 \end{cases}$$

$$\mu > \mu_0$$

$$\mu \neq \mu_0$$

4) Choose appropriate statistic.

$z, t, \dots$

5) Assume  $H_0 = \text{TRUE}$ .

set  $\mu = \mu_0$

6) Compute test statistic for observed data/sample.

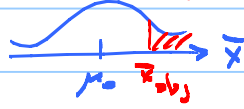
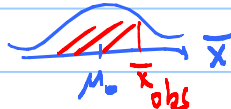
$$t_{obs} = \frac{\bar{x}_{obs} - \mu_0}{s/\sqrt{n}}$$

7) Find p-value, i.e. prob of getting a random test statistic more extreme (contrary to  $H_0$ ) than the observed one.

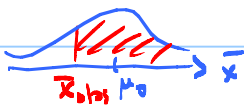
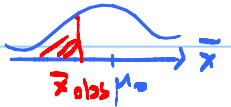
p-value =  $pr(\bar{X} < \bar{X}_{obs})$

$pr(\bar{X} > \bar{X}_{obs})$

... (see below)



...



...

$pr(t < t_{obs})$

$pr(t > t_{obs})$

...

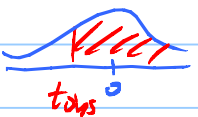
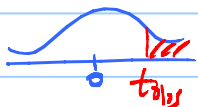
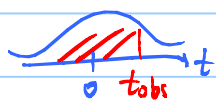
$prob(t > 2.91)$

mnemonic

left area

right area

sum of tail areas



8) If  $p\text{-value} < \alpha$ , reject  $H_0$  in favor of  $H_1$ ; else cannot reject  $H_0$  in favor of  $H_1$ .

Note: If  $p\text{-value} < \alpha$ , Then reject  $H_0$  in favor of  $H_1$ .

In English: There is evidence from data in favor of  $H_1$  ...  
(against  $H_0$ ).

Else, cannot reject  $H_0$  in favor of  $H_1$ .

(Not The same thing as "Accept  $H_0$ ")

In English: There is NO evidence from data in favor of  $H_1$  ...  
(against  $H_0$ ).

(Not The same thing as "There is evidence for  $H_0$ ")

We cannot Accept  $H_0$ , because we assumed it was true!

When  $p\text{-value}$  is large, There is simply no evidence from data for anything - not for  $H_1$ , not for  $H_0$ !

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As I said, There are lots of similarities between the C.I and the  $p\text{-value}$  approach, but the differences are very important.

For example there is no  $t^*$  (or  $z^*$ ) here. from Tables.

But There is  $t_{obs}$  (or  $z_{obs}$ ) from data

$$z^* \neq z_{obs}, t^* \neq t_{obs}$$

Now, let's repeat our example, but now following the general procedure.

Data says:  $n=64$ ,  $\bar{x}_{obs}=34.4$ ,  $S=1.1$ .

Does the data provide evidence to support  $\mu > 34$ ?

- 1) The param. of interest:  $\mu$  (dropping subscript  $x$ ).
- 2,3)  $H_0: \mu < 34$  (or  $\mu = 34$ )  $\mu_0 = 34$   
 $H_1: \mu > 34$
- Setting-up  $H_0/H_1$  is the hardest part of these problems. See below for guidance.

4) Appropriate test statistic:  $z, t$

5) Assume  $H_0 = T$ . (i.e. set  $\mu = 34$ )

6) Compute statistic assuming  $H_0 = \text{True}$  (i.e.  $\mu = \mu_0$ )  $t_{obs} = \frac{34.4 - 34}{1.1/\sqrt{64}} = 2.91$   
→ contrary to  $H_0$  (i.e. in the direction of  $H_1$ )

7)  $p\text{-value} = \text{prob}(\bar{x} > \bar{x}_{obs}) = \text{prob}(t > 2.91) \approx .0025$

8) Conclusion: At  $\alpha = .05$ ,  $p\text{-value} < \alpha$ .

Therefore,

Data provide sufficient evidence to reject  $H_0^{\mu < 34}$  in favor of  $H_1^{\mu > 34}$ .

"In English": Data provide suff. evidence in favor of  $\mu > 34$ .

At  $\alpha = 0.001$ ,  $p\text{-value} > \alpha$   $\mu < 34$

Therefore, we cannot reject  $H_0^{\mu < 34}$  in favor of  $H_1^{\mu > 34}$ .

"In English": there is no support for  $\mu > 34$

Note: This conclusion is NOT the same thing as "There is support for  $\mu < 34$ ". All we can say is that we cannot reject  $\mu < 34$ .

# Important.

The hardest part of hypothesis testing is setting-up  $H_0/H_1$  correctly. Here is some guidance:

Four ways I go about for deciding what  $H_0/H_1$  should be:

1) Don't assume what DATA are supposed to test.

The question asks "Does data provide evidence for claim X?" Meanwhile, the hypothesis testing procedure begins by assuming whatever you put under  $H_0$  is True. So, it makes no logical sense to assume X is true, even before data. So, put the complement/opposite of X under  $H_0$ .

2) Ask yourself what statement you should be left with if there is NO DATA at all. The answer to that question tells you what  $H_0$  should be. Then, the complement of that goes under  $H_1$ .

The data provide evidence for  $H_1$  (against  $H_0$ ), because of the way the whole procedure is set-up. Then, if the evidence is weak (eg when there is no data at all), then the procedure leaves you with  $H_0$ , as it should. In our example, if there is no data at all, then we should not reject the belief that  $\mu < 34$ , and so,  $H_0$  should be  $\mu < 34$ .

3) Some problems ask you to test some prior belief (i.e., some claim based on something other than data). Then that belief should go under  $H_0$ .

4) Another way of deciding on  $H_0/H_1$  will be discussed later, when we learn the meaning of alpha, and Type I and Type II errors.

Further comments:

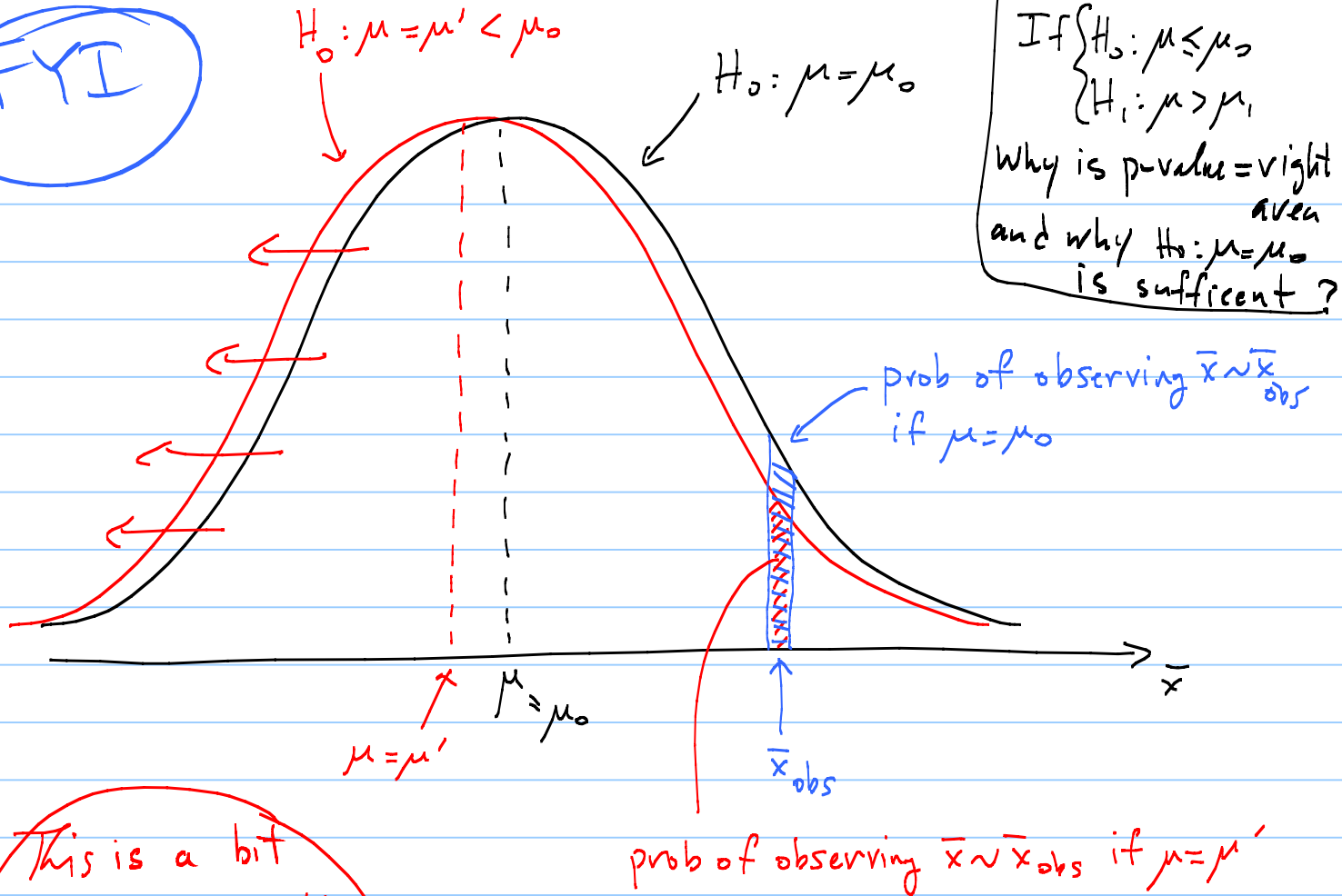
$H_0$  and  $H_1$  are statements about some pop. param, and so, they have no probability.

The p-value is the quantity that represents the evidence from data against  $H_0$ , in favor of  $H_1$ . But note that smaller p-value means more evidence against  $H_0$  (in favor of  $H_1$ ). This is so because we are giving the benefit of our doubt to  $H_0$ ; so, if  $H_0$  is true, and the prob of getting data more extreme than the observed data is large, then there is no evidence for rejecting  $H_0$ .

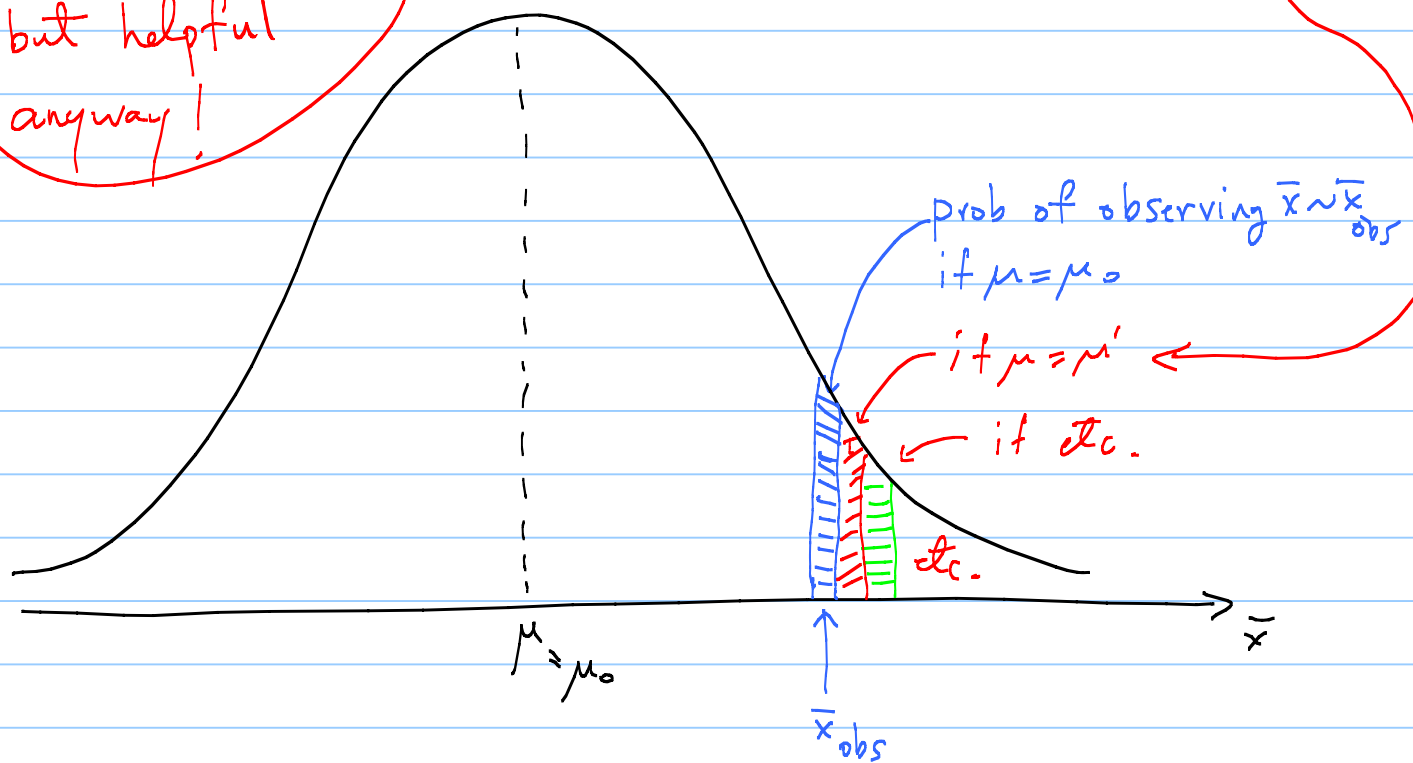
If we cannot reject  $H_0$  in favor of  $H_1$ , then we don't know anything. Not rejecting  $H_0$  is not the same thing as accepting  $H_0$ . Making that mistake of interpreting the lack of evidence for  $H_1$  as support for  $H_0$  is the source of much confusion in science.

In general, we cannot accept a belief about an unknown pop. parameter (eg.  $\mu < 34$ ). All we can do is either reject it, or not, based on evidence from data. And that evidence comes through the p-value; the mathematical way to see this is to note that the p-value is a conditional prob. i.e. it already assumes  $H_0$  is true.

FYI



This is a bit wrong/misleading but helpful anyway!



So, if  $H_0: \mu \leq \mu_0$ , then the p-value =  $\text{prob}(\bar{x} > \bar{x}_{obs} | \mu = 34)$   
 (i.e.  $H_1: \mu > \mu_0$ ) = right area, with  $\mu = 34$   
 mnemonic

### hw\_lect21\_1

Toothpaste tubes may be wasteful because there is always some amount of toothpaste that one cannot extract. To find out how much toothpaste is wasted, 5 discarded tubes are selected, cut open, and the amount of remaining toothpaste is recorded. The data are : 0.52, 0.65, 0.46, 0.50, 0.37 (in ounces). Is there evidence that the true average of the wasted toothpaste is less than 0.55 ounces? Apply the hypothesis testing procedure as follows:

- what is the pop. parameter being tested? Write the symbol for it, AND explain it in words.
  - Restate the question as "Does data provide evidence ---"
  - which of the following pairs of hypotheses is appropriate?
- Check the solns later to see the explanation/thinking.

$$H_0: \mu_x < 0.55$$

$$H_0: \mu_x > 0.55$$

$$H_0: \mu_x = 0.55$$

$$H_1: \mu_x > 0.55$$

$$H_1: \mu_x < 0.55$$

$$H_1: \mu_x \neq 0.55$$

- In our procedure we must assume  $H_0 = \text{True}$ . Assuming  $H_0 = T$ , what is the "worse" value that  $\mu_x$  can take? Hint: values in the direction of  $H_1$  are "worse" for  $H_0$ .
- Assuming the "worse-case" scenario of part d, compute the p-value. Hint: remember that the p-value measures evidence against (contrary to)  $H_0$ , or in favor of  $H_1$ ; Use Table VI.
- Is The p-value you have computed small (less than 0.05) or large (larger than 0.05)?
- Based on your answer to part f should you reject  $H_0$  in favor of  $H_1$ ?
- What is the conclusion (In English)?

### hw-lect 21-2

I Suppose you are asked if There is evidence That  $\mu_x > \bar{x}_{obs}$ ?

- Set-up The appropriate  $H_0/H_1$
- Compute The p-value.

II Suppose you are asked if There is evidence That  $\mu_x > \bar{x}_{obs} - 1.645 \frac{s}{\sqrt{n}}$ ?

For I: The right-hand side is The 95% LCB (which we are skipping).

- Set-up The appropriate  $H_0/H_1$

- Compute The p-value. Use  $\text{pr}(t > 1.645) = 0.05$ , ie.  $df = n-1 = \infty$

Look at The soln later to see the moral of This hw.