# Hypothesis Testing – Means and Proportions

8.1, 8.3

## Today's topics

#### **Hypothesis Testing**

- Definitions
- Testing for one mean
- p-value
- Testing for one proportion

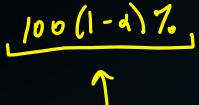
#### Statistics overview

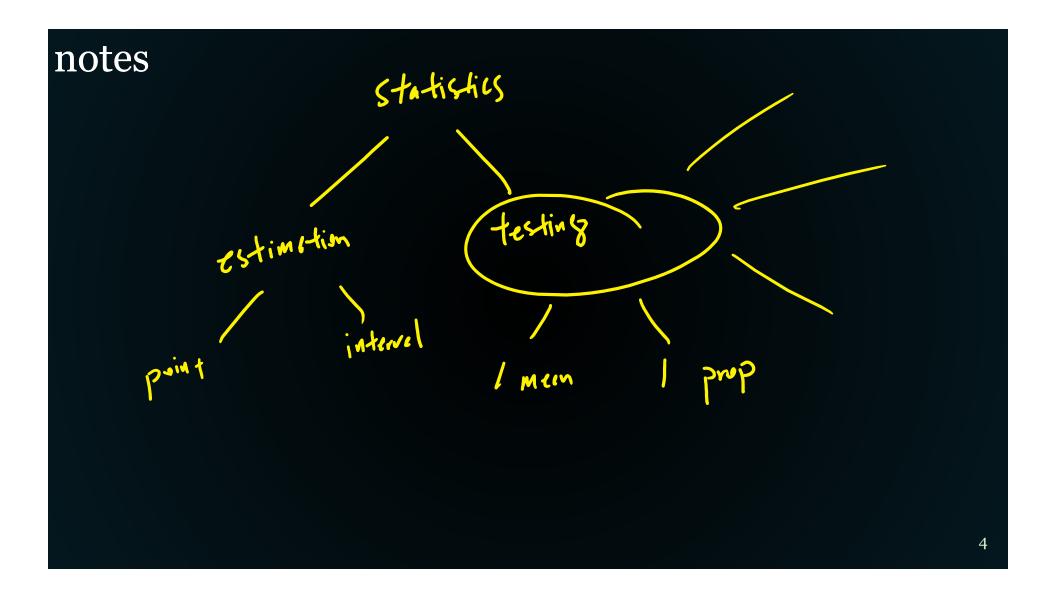
One goal in Statistics is to make *inferences* about populations based on samples taken from the population.

Previously, we estimated population parameters:

- Point estimates (MLE, MOM)
- Interval estimates (Confidence Intervals)







## Testing

Another way to do inference, is to **make a decision** about a parameter.

**Examples:** 



ctest claim

my Mazda 3 manual claims that it gets 35 highway mpg



Dustin's pudding packs actually contain 3.25 oz

#### Terms

- Null hypothesis, H<sub>0</sub>
- Alternative hypothesis,  $H_A$  or  $H_1$
- Type I error:
- Type II error
- Simple hypothesis
- Compound hypothesis

$$H_A: \mu < 35$$

# Null and Alternative Hypotheses

Say an experimenter wants to test the plausibility of the statement  $\mu = \mu_0$ .

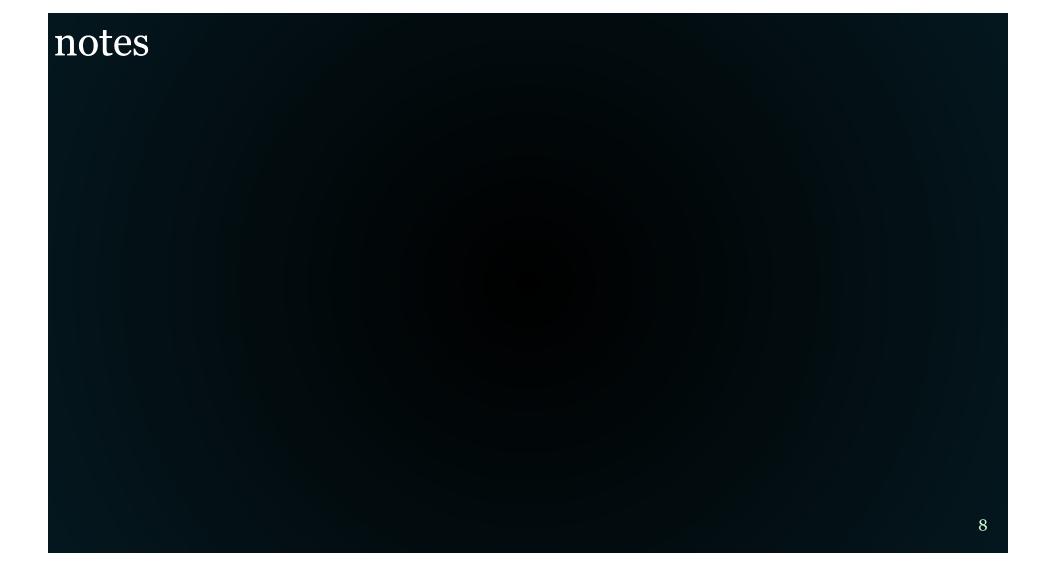
We can formally describe this as a null hypothesis.

- $\blacksquare$   $H_0: \mu = \mu_0$  + mean from nul
- The word "hypothesis" indicates that we will be testing this statement (with data).

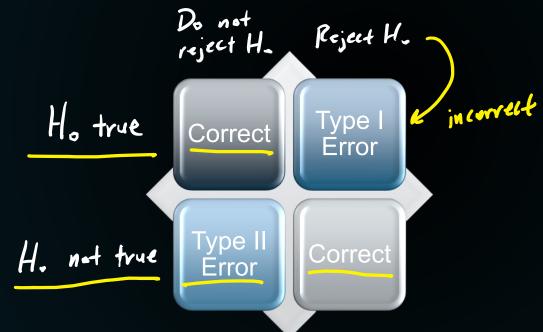
We will associate the null hypothesis with a different one that we are testing 'for', called the "alternative hypothesis".

$$\blacksquare \quad \mathbf{H}_{\mathsf{A}} \colon \ \mu \neq \mu_0 \quad \quad \mathsf{O}$$

$$\mathbf{H_A}$$
:  $\mu > \mu_0$  or  $\mathbf{H_A}$ :  $\mu < \mu_0$ 



## Type 1 and Type 2 Error

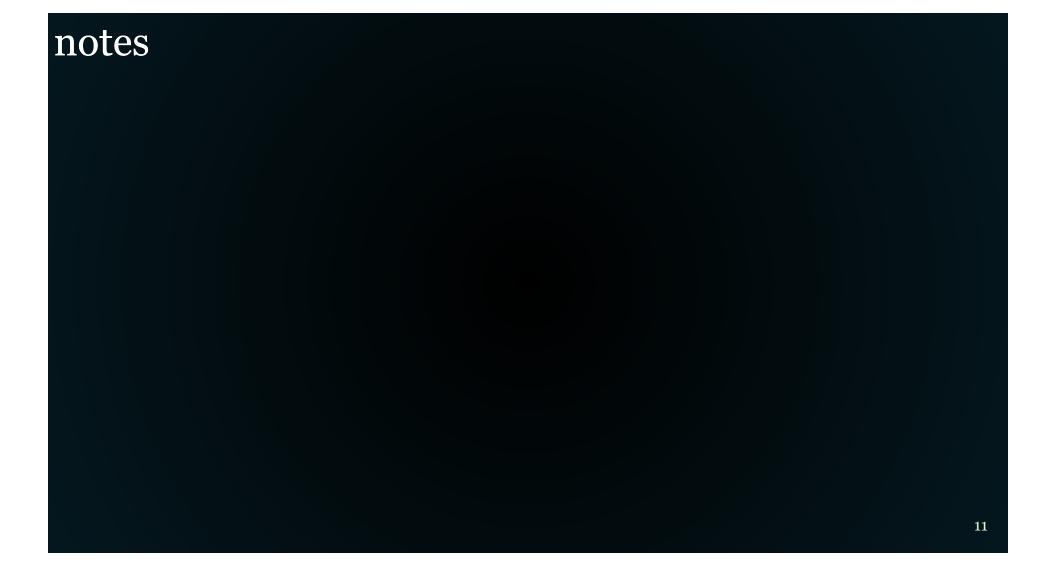


# Hypothesis Test Example – "Compound" H<sub>A</sub>

Perdont university claims that students at their school are above average intelligence. A random sample of thirty students IQ scores have a mean score of 102.5.

Suppose the mean population IQ score is 100 with a standard deviation of 15. Is there sufficient evidence to support this claim?





### Hypothesis Test Example – "Simple" H<sub>A</sub>

Example 8.1-1

Let X equal the breaking strength of a steel bar. If the bar is manufactured by process I, X is N(50,36), i.e., X is normally distributed with  $\mu = 50$  and  $\sigma^2 = 36$ . It is hoped that if process II (a new process) is used, X will be N(55,36). Given a large number of steel bars manufactured by process II, how could we test whether the five-unit increase in the mean breaking strength was realized?

$$H_0: \mu = 50$$
  $H_0: \mu \le 50$ 

$$H_1: \mu = 55$$
  $H_2: \mu = 55$ 

We want to set up a "rule" to determine whether to stick with  $H_0$  or not. This rule will lead to a decision about what to do with  $H_0$ .

Partition sample space into 2 parts: C and C'.

If 
$$(x_1, x_2,...x_n) \in C$$
, reject  $H_0$   
If  $(x_1, x_2,...x_n) \in C'$ , do not reject  $H_0$ 

e.g. 
$$\mathbf{x} = \{64.4, 54.7, 57.2, 61.6, 51.3\}$$
 or  $\mathbf{x} = \{51.2, 54.7, 47.2, 51.6, 46.3\}$ 

We often partition the sample space in terms of values of a statistic called a **test statistic**.



Often, we partition the sample space based on the value of a statistic called the **test statistic**.

One common example is  $\overline{X}$  (for testing the mean).

We might want to reject the null hypothesis if the sample average is larger or smaller than a certain number. E.g.  $\overline{X} > 53$ .

C is referred to as the rejection region, or the critical region.



i.e. C = {(x, x2, ... Xx): x > 53}

p value

The plausibility of a null hypothesis can be measured with a **p-value**, which is a number between 0 and 1.

- A p-value is sometimes referred to as the <u>observed level of</u> significance
- The smaller the p-value, the less plausible  $H_0$  is.

#### **Definition of a p-value:**

"Probability of observing data at least as extreme as the observed sample given that H<sub>o</sub> is true."



# p-value illustration

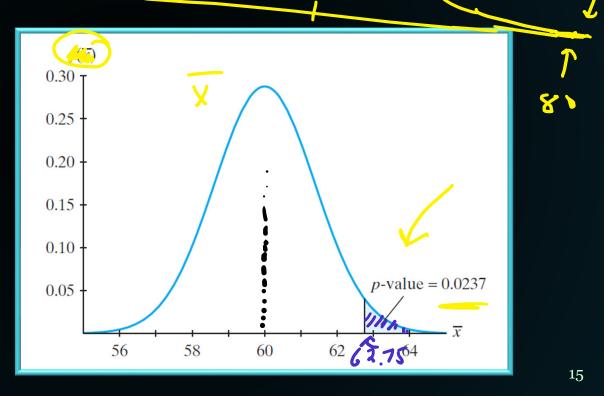
$$\rightarrow$$
 **H**<sub>0</sub>:  $\mu = 60$  vs

$$\rightarrow \mathbf{H_0}: \mu = 60$$

$$\rightarrow \mathbf{H_A}: \mu > 60$$

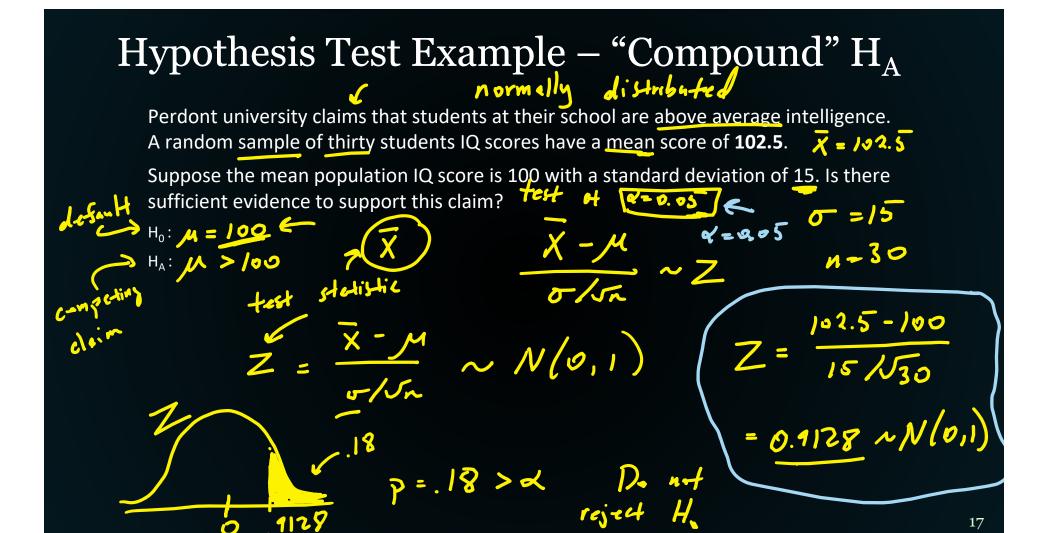
Suppose 
$$\bar{x} \neq 62.75$$

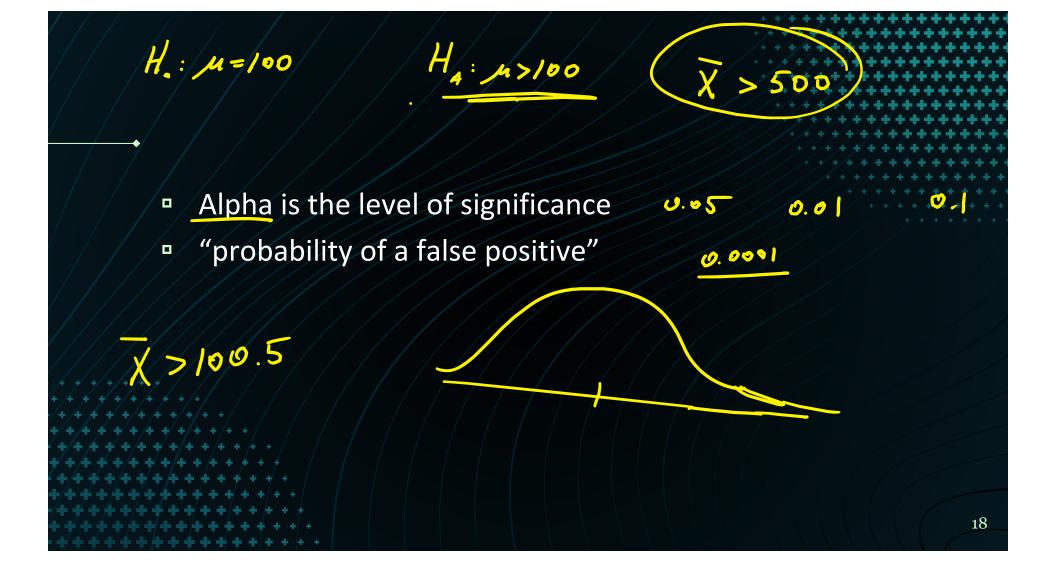




## Hypothesis Testing (Steps)

- $\rightarrow$  1. Formulate H<sub>0</sub> and H<sub>A</sub> (based on the scenario)
- $\rightarrow$  2. Identify a test statistic to use and its distribution under  $H_0$
- 3. Evaluate the test statistic
- $\rightarrow$  4. Calculate a p-value, compare to  $\alpha$ .
  - 5. Make a decision
  - if  $p < \alpha$ , reject  $H_0$ . Otherwise, (if  $p > \alpha$ ), do not reject  $H_0$ .
  - 6. State conclusion in the context of the original question.
    - "There is/isn't enough evidence to show that..."





## Two ways to perform a hypothesis test

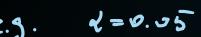
given

1. Calculate a p-value and compare to significance level

is P > a, DNR H.

2. Define a rejection region (RR) and see if sample falls in

RR. (also known a critical region)



l-sided

#### Table 8.1-1 Tests of hypotheses about one mean, variance known

$$H_0$$
  $H_1$   $\mu = \mu_0$   $\mu > \mu_0$   $\mu > \mu_0$   $\mu = \mu_0$   $\mu < \mu_0$   $\mu < \mu_0$   $\mu = \mu_0$   $\mu < \mu_0$   $\mu < \mu_0$   $\mu = \mu_0$   $\mu < \mu_0$   $\mu = \mu_0$   $\mu < \mu_0$   $\mu = \mu_0$   $\mu \neq \mu_0$   $\mu$ 

$$Z = \frac{\overline{X} - \mu_0}{\sqrt{\sigma^2/n}} = \frac{\overline{X} - \mu_0}{\sigma/\sqrt{n}}$$

#### Table 8.1-2 Tests of hypotheses for one mean, variance unknown

| $H_0$         | $H_1$            | Critical Region   |
|---------------|------------------|---|
| $\mu = \mu_0$ | $\mu > \mu_0$    | $t \ge t_{\alpha}(n-1)$ or $\overline{x} \ge \mu_0 + t_{\alpha}(n-1)s/\sqrt{n}$         |
| $\mu = \mu_0$ | $\mu < \mu_0$    | $t \le -t_{\alpha}(n-1)$ or $\overline{x} \le \mu_0 - t_{\alpha}(n-1)s/\sqrt{n}$        |
| $\mu = \mu_0$ | $\mu \neq \mu_0$ | $ t  \ge t_{\alpha/2}(n-1)$ or $ \overline{x} - \mu_0  \ge t_{\alpha/2}(n-1)s/\sqrt{n}$ |

$$T = \frac{\overline{X} - \mu}{\sqrt{S^2/n}} = \frac{\overline{X} - \mu}{S/\sqrt{n}}.$$

## Example

A machine that pours beer is supposed to fill buckets with a mean of 3.0 gallons. A sample of 13 buckets gives a sample mean of 2.879 gallons and s = 0.325. Perform a hypothesis test at  $\alpha$ = 0.05 to see if this machine is

accurately doing its job.

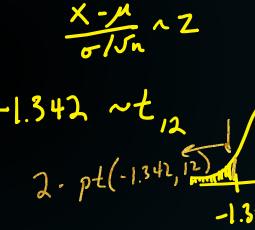
$$H_0: \mu = 3.0$$
 $H_A: \mu \neq 3.0$ 

2-sided test

p-value = 204

Decision: > <

>2 DNR H



-1.342 0 1.342

Conclusion: There is not enough evidence to suggest that the machine is not accurately doing its job

## Example



A HAL-8000 machine that pours beer is supposed to fill buckets with a mean of 3.0 gallons. Dave takes a sample of 13 buckets and finds a sample mean of 2.879. Suppose the true standard deviation of these machines is 0.2 gallons. Perform a hypothesis test at  $\alpha$ = 0.05 to see if this machine is

underfilling.

H<sub>0</sub>: 
$$M = 3.0$$

H<sub>A</sub>:  $M = 3.0$ 
 $Z = 1$ 
 $Z = 2.879 = 3.0$ 
 $Z = 2.879 = -2.18$ 

p-value = 0.015 < 0.05

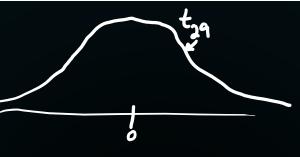
Decision:  $P < \alpha$   $Z = 3.0$ 

Decision:  $P < \alpha$   $Z = 3.0$ 
 $Z = 3$ 

Conclusion: There is enough evidence to suggest that this machine is underfilling. (H. Ind.)

$$T = \frac{\overline{X} - \mu}{\sqrt{S^2/n}} = \frac{\overline{X} - \mu}{S/\sqrt{n}}.$$

#### Example



normally distributed

Nick Fury claims that the (true) mean number of push-ups his superheroes can do is at least 40.0. A random sample of 30 superheroes gives  $\bar{x} =$ 38.518 pushups, s = 2.299. Perform a hypothesis test at  $\alpha = 0.01$  to determine if this is true (or if they can't really make it to 40 pushups).

$$H_A$$
:  $M < 40$ 

$$t = \frac{38.518 - 40^{\circ} t_{29}}{2.233 / \sqrt{30}}$$

t = -3.53t = 39.518 - 40 ~ t<sub>29</sub> 2219/130 (4) p-value: 0.0007 < 0.01 -3.53

Conclusion: There is significant evidence to suggest that they can NOT do 40

| Table 8.3-1 Tests of hypotheses for one proportion |              |  |  |  |
|--|--------------|--|--|--|
| $H_0$  | $H_1$        | Critical Region  |  |  |
| $p = p_0$  | $p > p_0$    | $z = \frac{y/n - p_0}{\sqrt{p_0(1 - p_0)/n}} \ge z_{\alpha}$       |  |  |
| $p = p_0$  | $p < p_0$    | $z = \frac{y/n - p_0}{\sqrt{p_0(1 - p_0)/n}} \le -z_{\alpha}$      |  |  |
| $p = p_0$  | $p \neq p_0$ | $ z  = \frac{ y/n - p_0 }{\sqrt{p_0(1 - p_0)/n}} \ge z_{\alpha/2}$ |  |  |

$$Z = \frac{Y/n - p_0}{\sqrt{p_0(1 - p_0)/n}}$$

## Hypothesis test for proportions

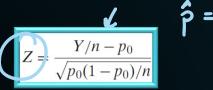
Aladdin is a frequent flyer. He thinks he gets security-screened more than normal at the magic-carpet-port. Assume security randomly screens 10% of all people (so he should be screened 10% of the time). In the past few years, he has been (randomly?) selected 16 out of 100 times.

Perform a hypothesis test at 0.05 significance to see if the screening process is random, or biased towards screening him more.





## Hypothesis test for proportions



Assume security randomly screens 10% of all people (so he should be screened 10% of the time). Aladdin has been selected 16 out of 100 times. Is process biased towards screening him

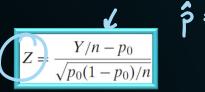
more? Test at  $\alpha$ =0.05.

Conclusion:

Test statistic:  $Z = \frac{1}{\sqrt{P \cdot UP \cdot}} = \frac{16 - 1}{\sqrt{100}}$ p-value:  $0.0217 = \sqrt{P \cdot UP \cdot} = \sqrt{100}$ Decision:  $P < A = \sqrt{25} cut H_{\bullet}$ Conclusion:

There is significant evidence to suggest that the

Screening process is biased towards screening Aladdin more



0

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