



Hypothesis testing, p-values, t-tests (part 1)

With your neighbor, come up with a graph of a cumulative distribution function for a fair 6-sided die.

Kind of distribution: uniform

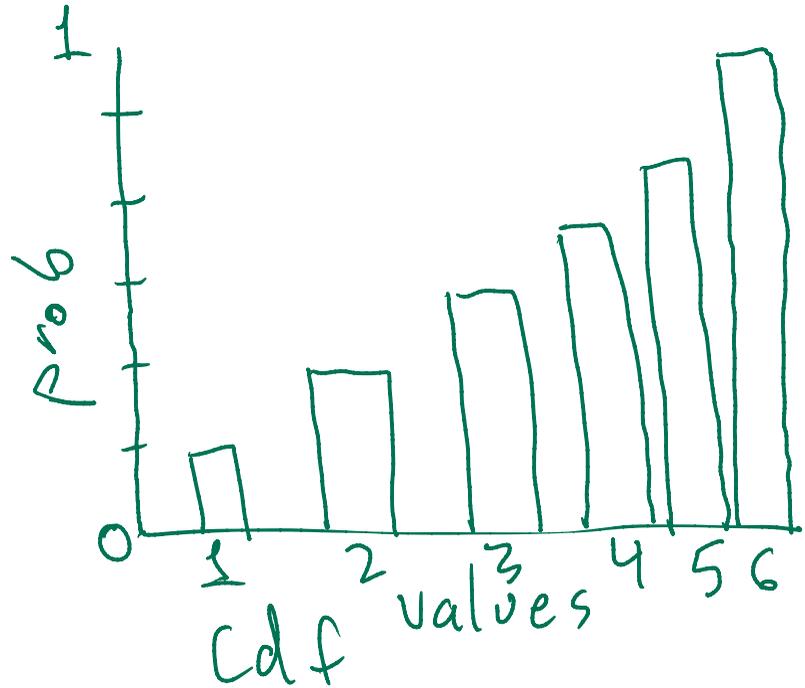
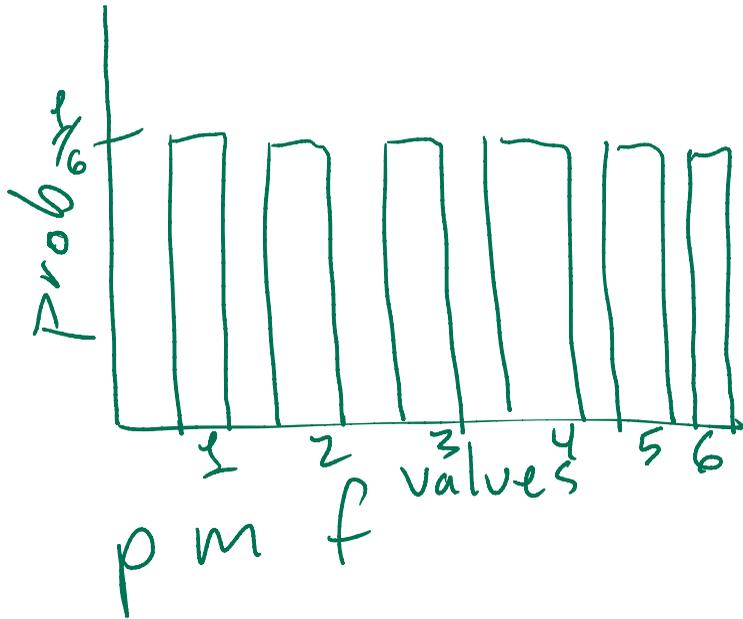
discrete vs. real: discrete

enumerable options $\{1, 2, 3, 4, 5, 6\}$

pmf vs. pdf: pmf (because X is discrete)

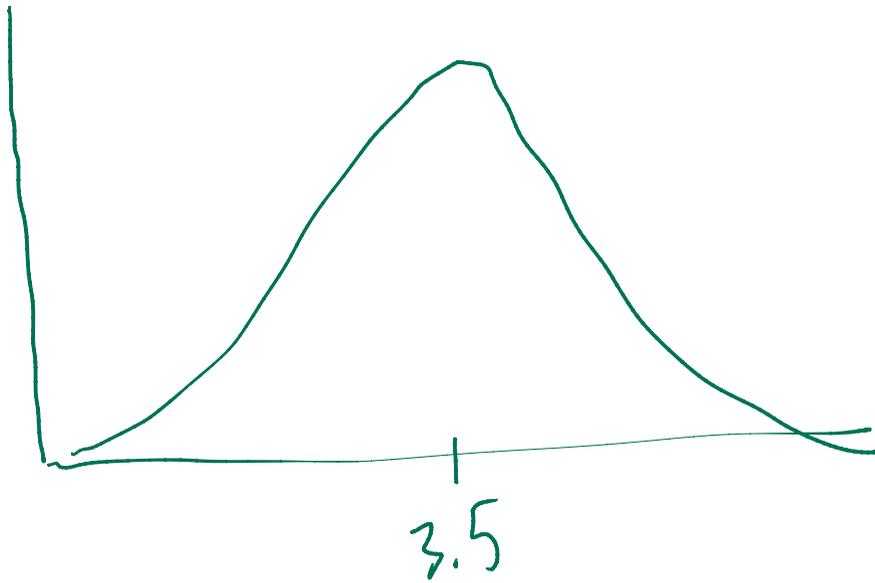
Cdf for a die

↳ prob that r.v. is \leq a target value



Central limit theorem:

↳ plot the means of samples of ind.
r. v.s (same properties) \rightarrow normal dist.



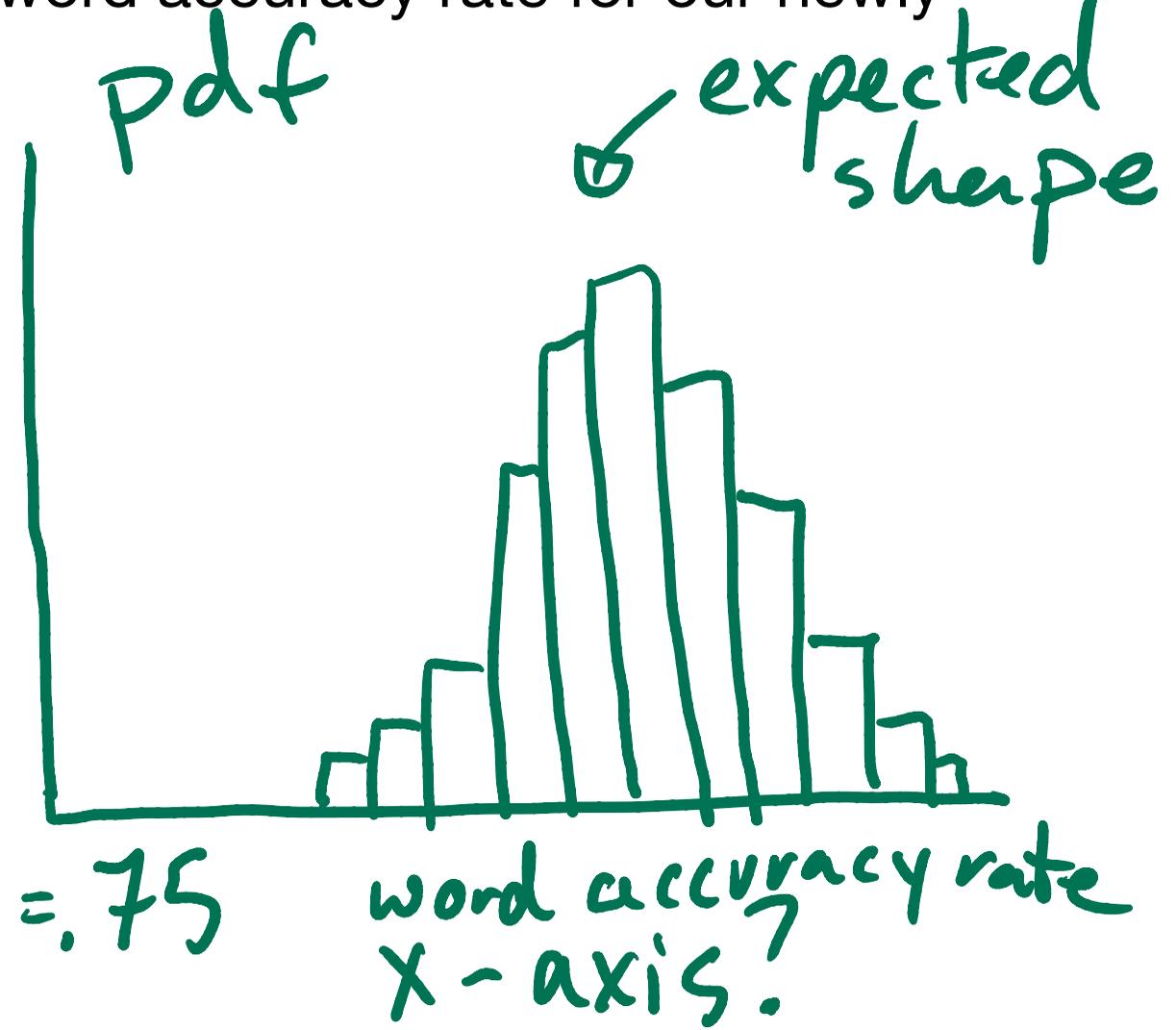
ICA Question 1: central limit theorem

Say that we are trying to determine the word accuracy rate for our newly developed voice assistant technology.

Samples

| w_1 | ... | w_n | |
|-------|-----|-------|-----|
| ✓ | ✓ | ✓ | 1 |
| ✓ | x | ✓ | .75 |
| ✓ | x | x | .5 |
| ✓ | x | ✓ | .5 |
| ✓ | ✓ | ✓ | 1 |

$p(w_{\text{correct}}) = .75$

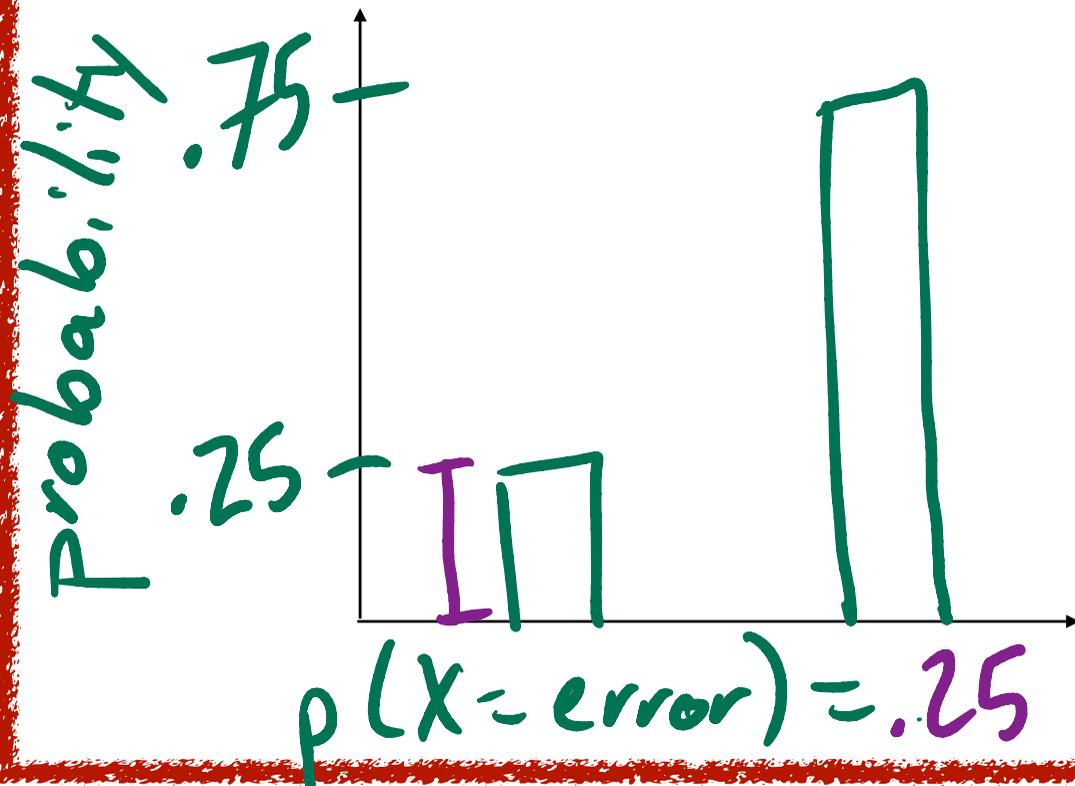


ICA Question 2: pmf vs. pdf

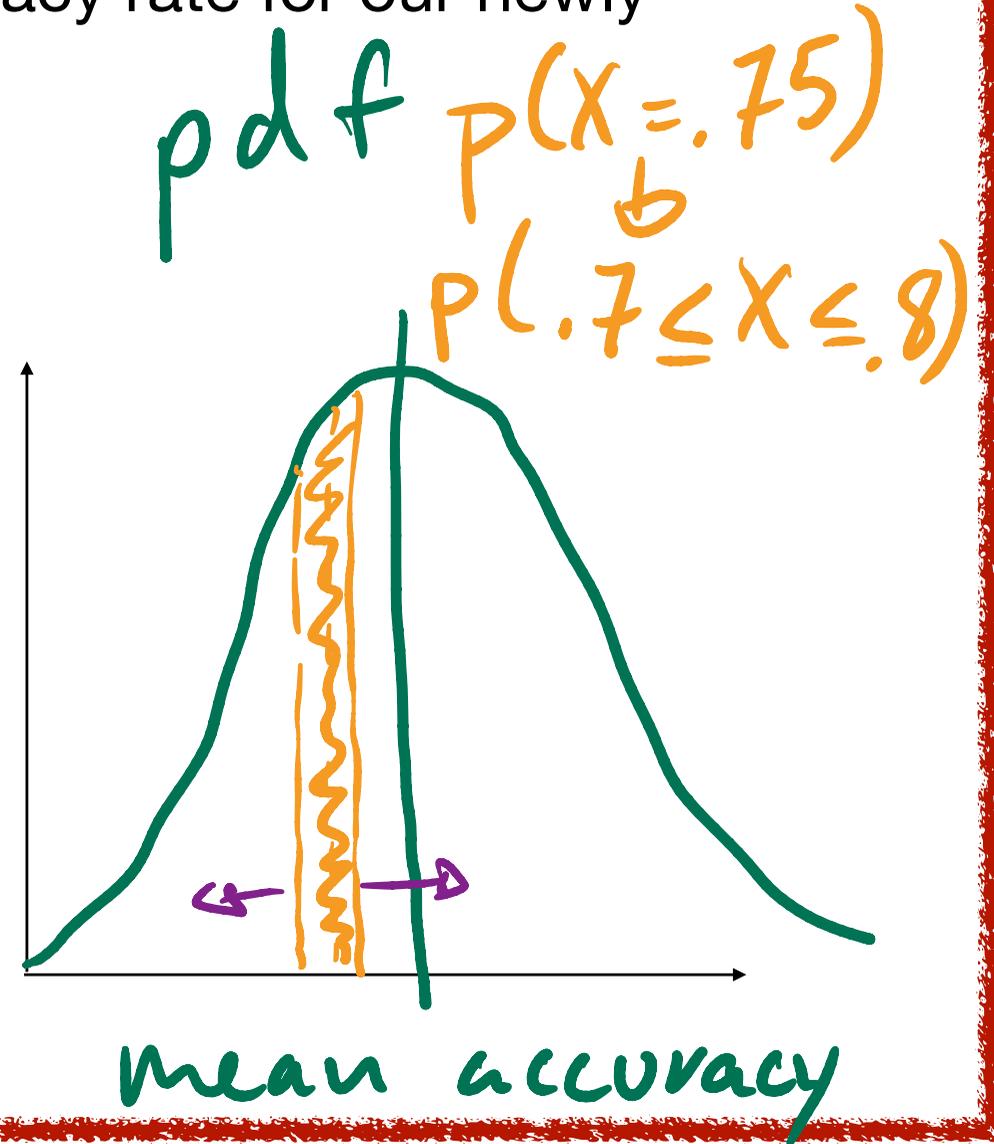
$$P(0 \leq X \leq 1) = 1$$

Say that we are trying to determine the word accuracy rate for our newly developed voice assistant technology.

pmf



probability per unit x



Wait, why is the probability of a value for for a real-valued random variable 0?

- The practical part:

$$P(x = .75) \neq 0$$

$$P(x = .751) \neq 0$$

$$P(x = .7513) \neq 0$$

$$P(.745 \leq x \leq .755)$$

because infinite #s,

$$\sum_{\phi} \#s = \text{infinity}$$

non-zero, positive

- For probability **density** functions, we care about **area** for probability, and that for probability **mass** functions, we care about **height** for probability (y-axis)
- for pdfs, y-axis is probability per x-unit ("probability density")

for a non-zero prob. in practice

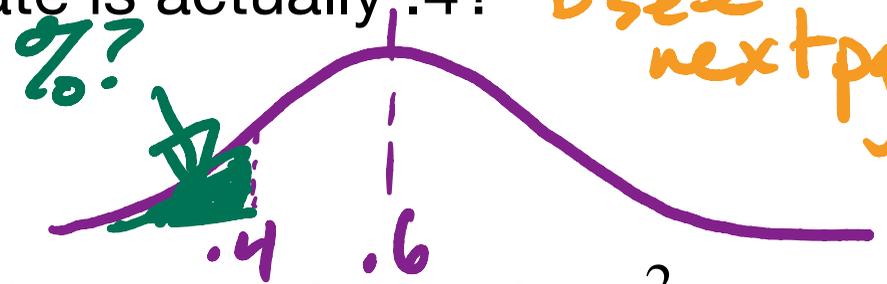
ICA Question 3: cdf vs. ppf

cdf → in: real val out: %
ppf: in: % out: real val

Say that we are trying to determine the word accuracy rate for our newly developed voice assistant technology.

↳ used the cent. lin. the → normal dist.

Given a μ of .6 and a σ^2 of 0.05, what is the chance that the true word accuracy rate is actually .4? ** → see next pg*

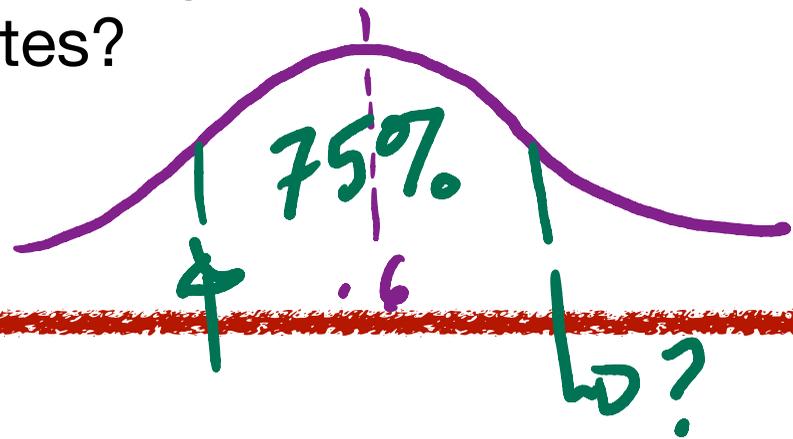


cdf
in: .4

$\sigma = .224$

$\text{norm.cdf}(.4, \mu, \sigma)$

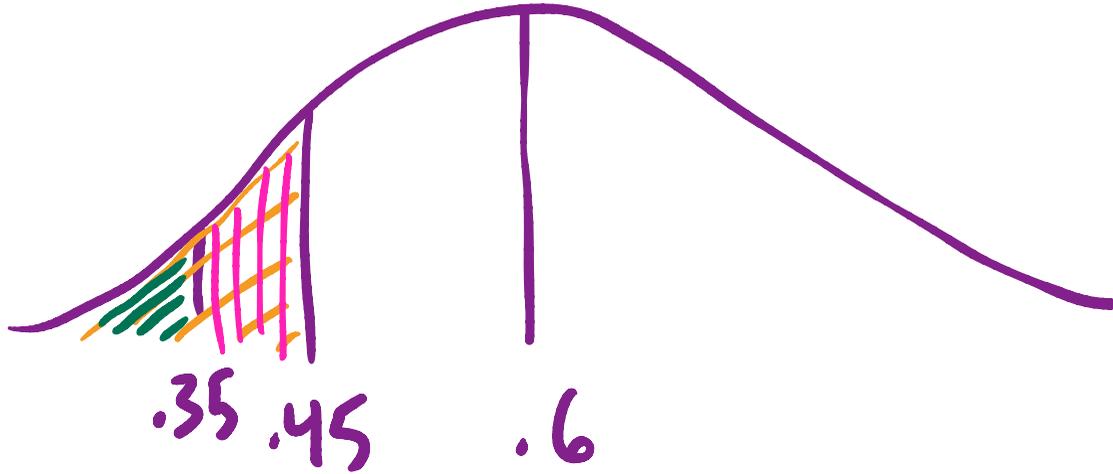
Given a μ of .6 and a σ^2 of 0.05, what is the **lower bound** for the true word accuracy rate if we want to claim that we are in the top 25% of possible accuracy rates?



$\text{norm.ppf}(1-.25, \mu, \sigma)$

"exactly" .4

$$\underline{cdf(.45, \mu, \sigma)} - \underline{cdf(.35, \mu, \sigma)}$$



pdf(.4) \rightarrow theoretically, yes

Hypothesis

- A **hypothesis** is a tentative assumption made in order to draw out and test its logical or [empirical](#) consequences
- (Merriam-Webster)

↳ bring this to math land, testing assumptions about distributions.

Hypothesis testing

- We'll be starting with a **question**

- Is there a change in student test scores based on whether or not they listen to music beforehand?

- Next, we'll need to describe some **observations**

M - students who listened to music + test scores

S - students who didn't + test scores

- Then, we'll write down the **hypothesis** being tested

$$H_1: \mu_M \neq \mu_S$$

The null hypothesis

- The **null hypothesis**— H_0 is the hypothesis that there is no difference between the observed groups
- For example, given the question:
 - Is there a change in student test scores based on whether or not they listen to music beforehand?
 - with the hypothesis: $H_1 : \mu_{music} \neq \mu_{nomusic}$
 - the null hypothesis is $H_0 : \mu_{music} = \mu_{nomusic}$

The null hypothesis

- The **null hypothesis**— H_0 is the hypothesis that there is no difference between the observed groups
- For example, given the question:
 - Do students who eat strawberries for breakfast have higher test scores than students who don't?

- with the hypothesis: $H_1: \mu_{\text{strawbs}} > \mu_{\text{no strawbs}}$

- the null hypothesis is: $H_0: \mu_{\text{strawbs}} = \mu_{\text{no strawbs}}$

ICA Question 4: hypotheses

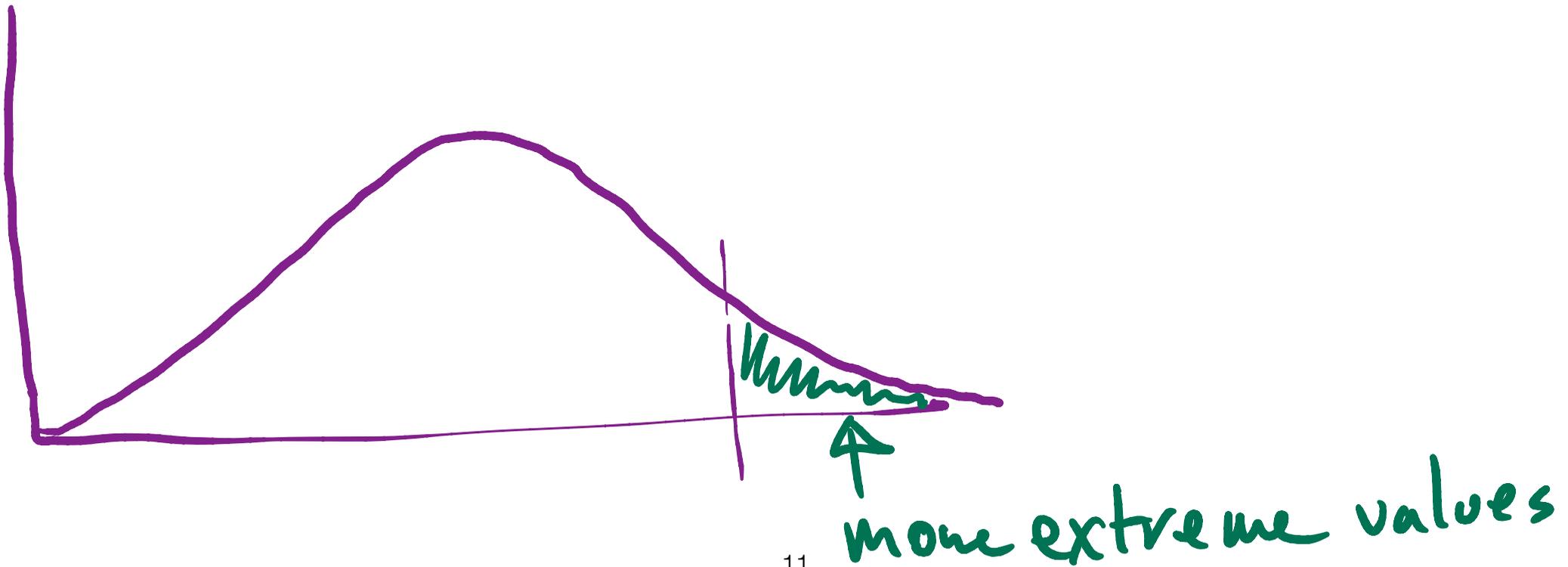
Come up with a **question**, what **observations** you'd need to answer it, and what your **hypothesis** is.

| <u>questions</u> | <u>observations</u> | <u>hypo</u> s |
|--|-----------------------------|-----------------------------|
| med. cure headache? | headaches w/ med w/o med | $\mu_{med} < \mu_{no\ med}$ |
| gas mileage w/ gas types? | cars w/ prem cars v/ reg | $\mu_{prem} > \mu_{reg}$ |
| ↳ issues to think about: diff. car types, gas source | | |

P-values

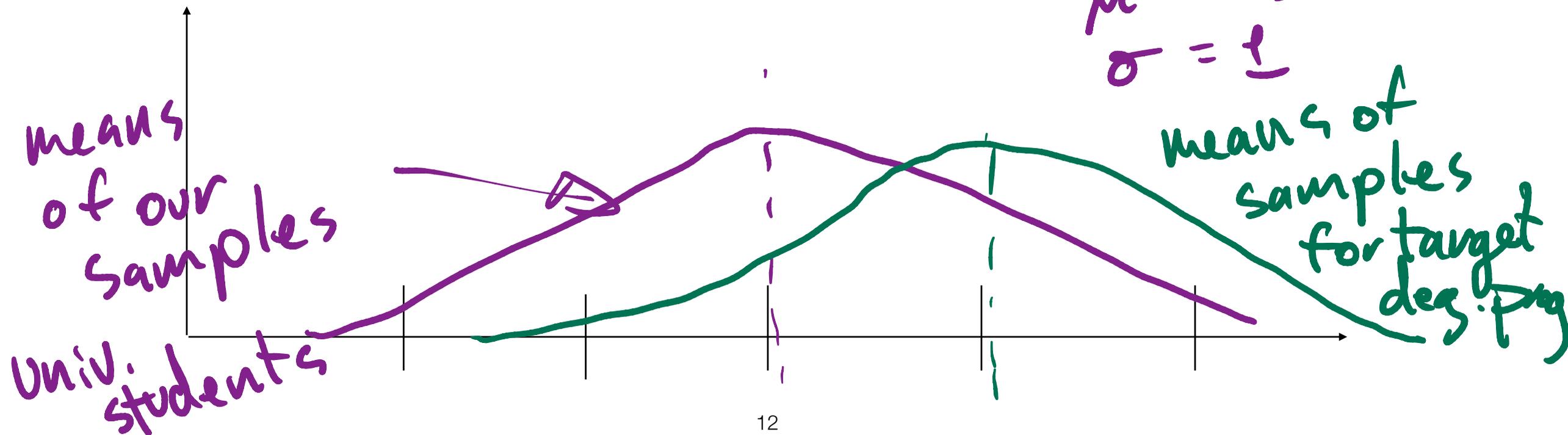
- A **p-value** is the probability of observing test results that are at least as extreme as the results that were actually observed.

↳ ⚠ we don't use p-values to describe individuals!



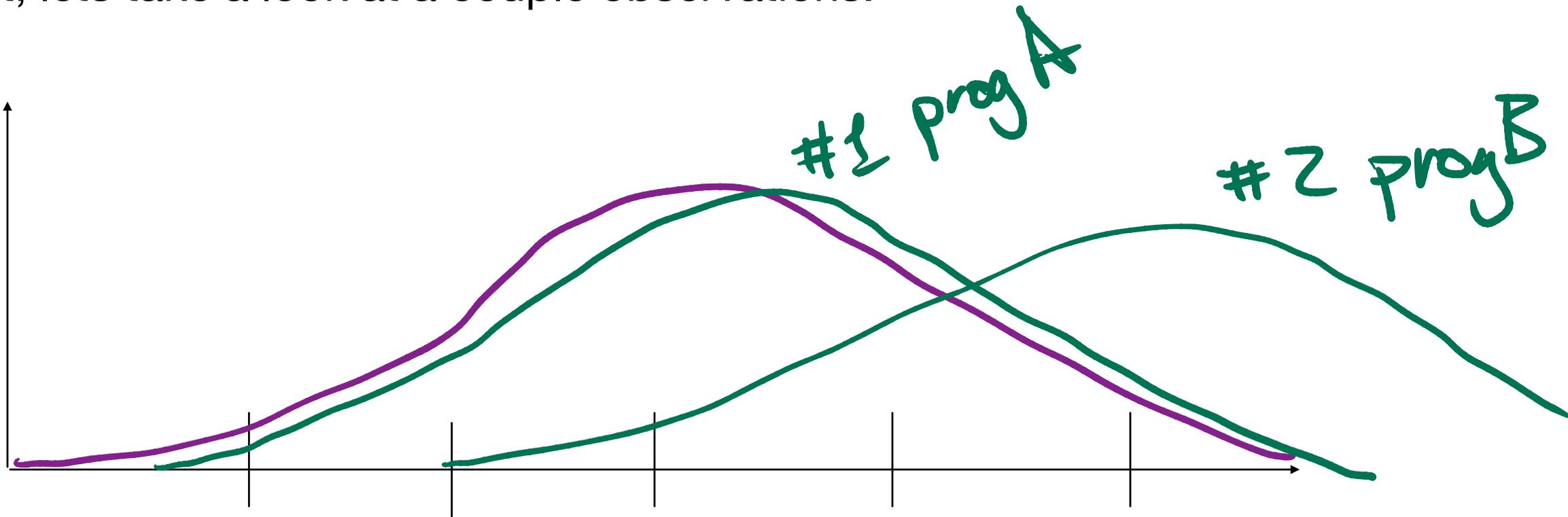
P-values

- Say that I want to know if a population of students in a certain degree program has a mean age that is significantly different than the mean ages of students in the university as a whole.
- First, we'll rely on the central lim. th., to build a distribution of mean ages of students in the university as a whole.



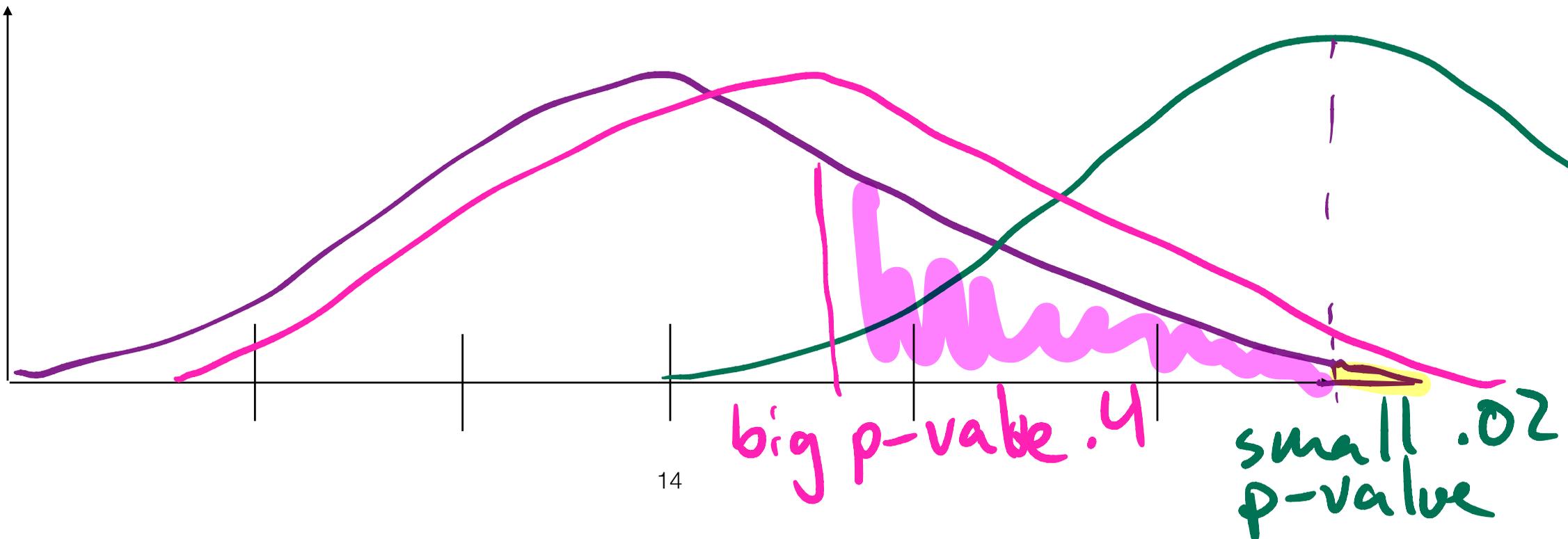
P-values

- Say that I want to know if a population of students in a certain degree program has a mean age that is significantly different than the mean ages of students in the university as a whole.
- Next, lets take a look at a couple observations:



P-values

- Say that I want to know if a population of students in a certain degree program has a mean age that is significantly different than the mean ages of students in the university as a whole.
- A larger p-value means that we are **more likely** to observe something that is **at least as extreme** as what we have observed.



P-values

- Say that I want to know if a population of students in a certain degree program has a mean age that is significantly different than the mean ages of students in the university as a whole.
- What do we need to calculate a p-value?
 - null hypothesis
 - test statistic
 - data / observations

test statistics

- Remember: overall goal is to be able to answer the question "is what I have observed meaningfully different than what I expect?" (vs. just due to random chance)

- We want to know if a coin is fair.

- null hypothesis $H_0: P(\text{heads}) = 0.5 = P(\text{tails})$

- test statistic count the heads

- data - counts of heads in the sample

test statistics

- Remember: overall goal is to be able to answer the question "is what I have observed meaningfully different than what I expect?" (vs. just due to random chance)
- We want to know if a population has a different mean age than another population

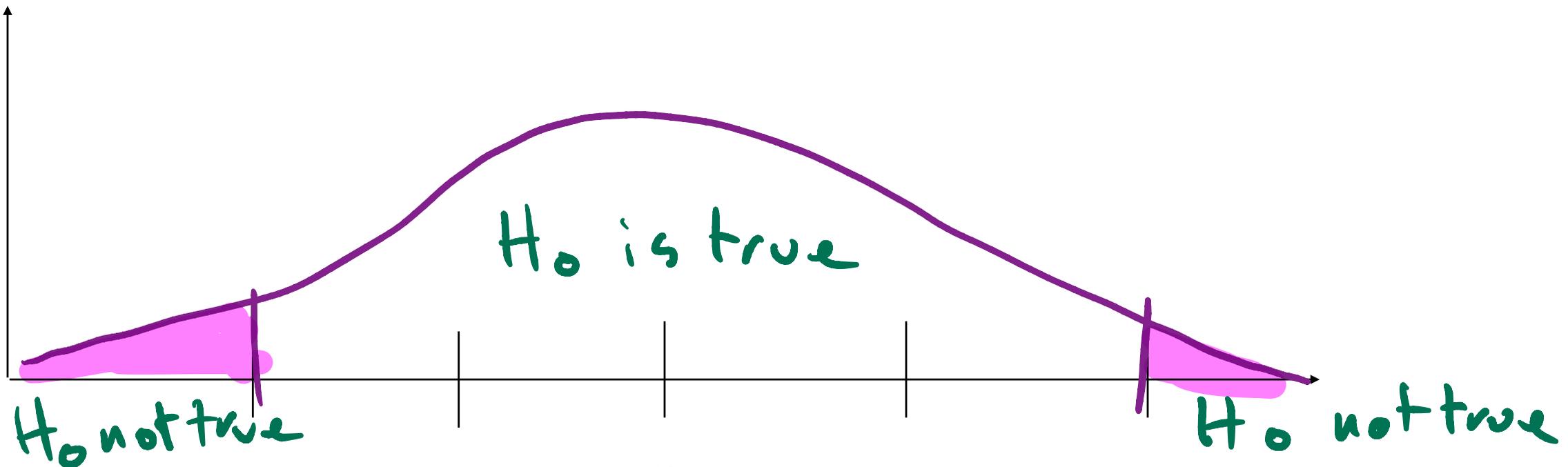
- null hypothesis $H_0: \mu_{p_1} = \mu_{p_2}$

- test statistic t -statistic (from t -test)

- data
ages of people in p_1
ages of people in p_2

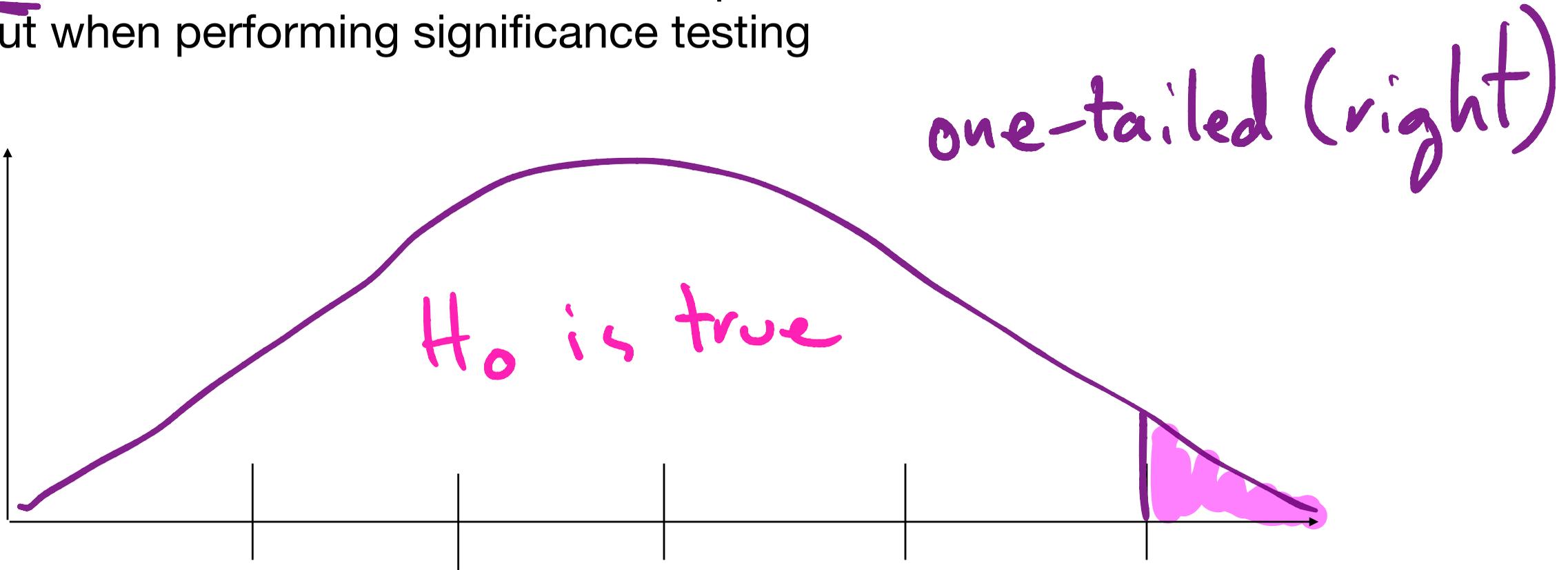
test statistics & tails

- We want to know if a population has a **different** mean age than another population
- **one** vs. **two** tailed tests refer to which part of the distribution we care about when performing significance testing



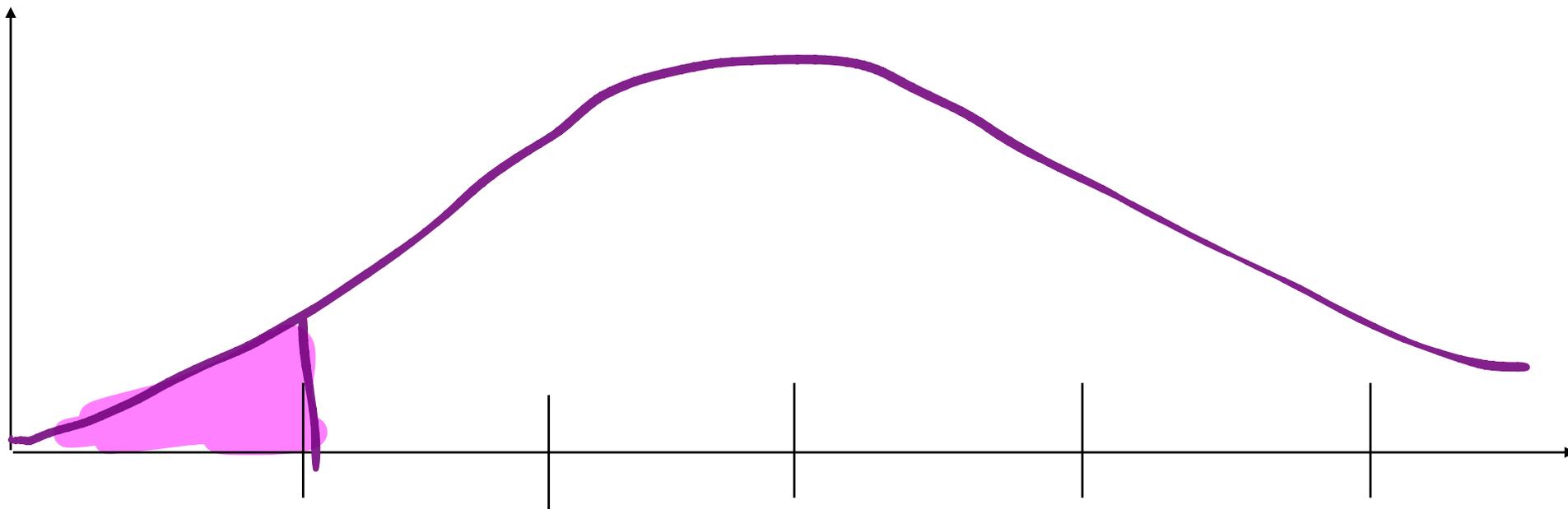
test statistics & tails

- We want to know if a population has a **larger** mean age than another population
- **one** vs. **two** tailed tests refer to which part of the distribution we care about when performing significance testing



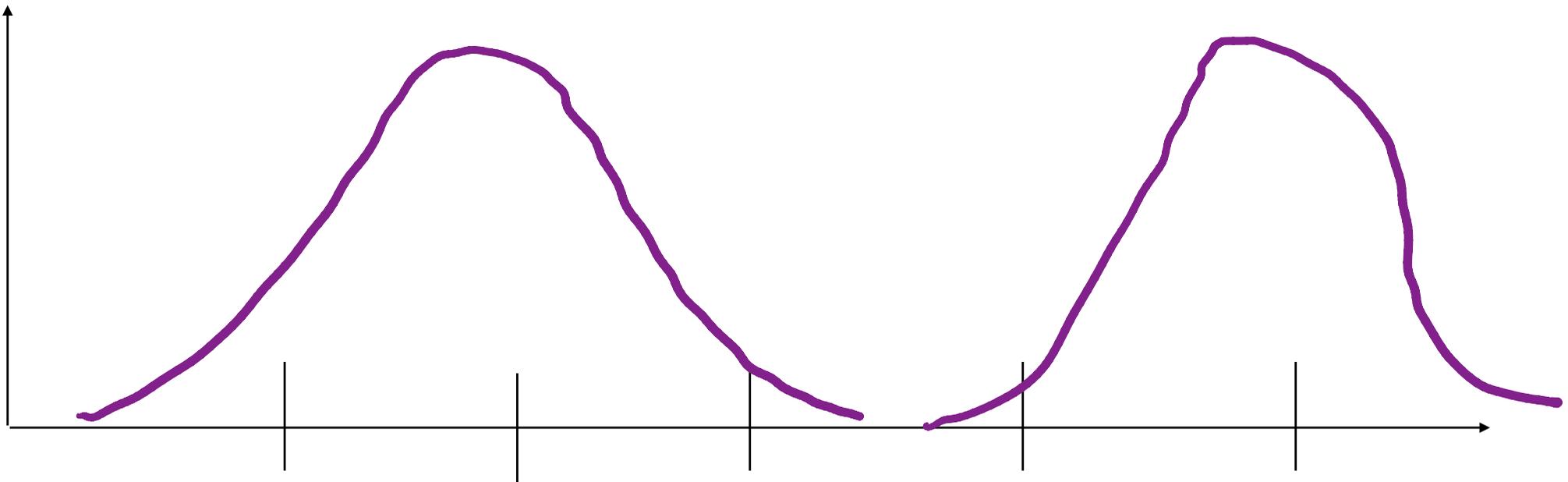
test statistics & tails

- We want to know if a population has a **smaller** mean age than another population
- **one** vs. **two** tailed tests refer to which part of the distribution we care about when performing significance testing



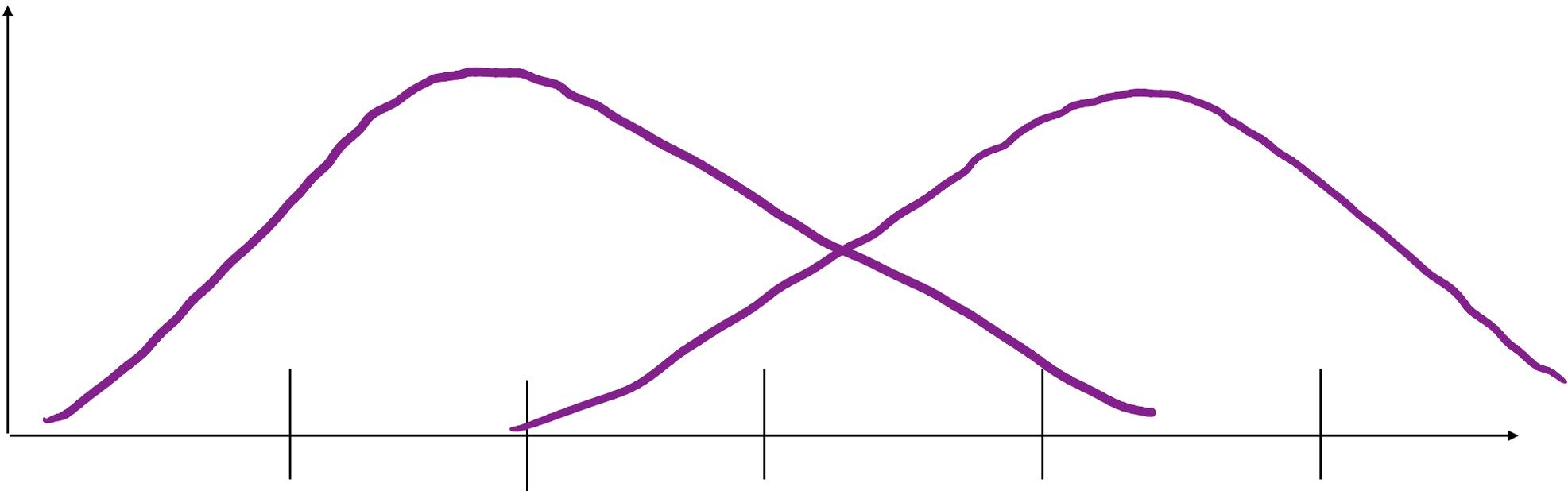
t-tests

- Student's t-test is the name of the test statistic that we'll use when we're trying to compare two continuous probability distributions that are normally distributed.



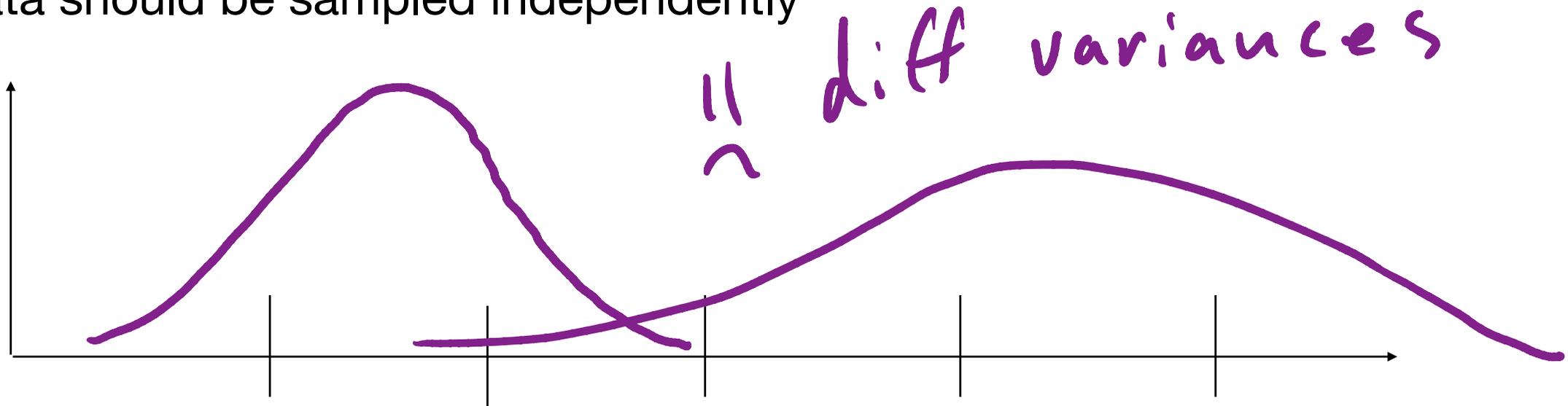
t-tests

- **Student's t-test** is the name of the test statistic that we'll use when we're trying to compare two continuous probability distributions that are normally distributed.



t-tests

- Before we go wild with t-tests on everything, there are a few requirements!
 - distributions should be normal
 - the two populations should have the same variance
 - data should be sampled independently



Future-you

- On Monday:
 - Actually calculating t-tests (and p-values)
↳ one ques on HW 7
 - errors
> one ques on HW 7
 - bias
 - mis-using p-values "harking"

Schedule

ICA : passcode : "cookie"

Turn in ICA 17 on Canvas (make sure that this is submitted by 2pm!)

HW 7 - available on the course website/canvas now. Due April 3rd. You will need some material from lecture on Monday!

| Mon | Tue | Wed | Thu | Fri | Sat | Sun |
|--|---|--|--|-----|-----|-----------------|
| March 21st Lecture 16 - normal distributions | Felix OH Calendly HW 6 due @ 11:59pm | Felix OH Calendly | Felix OH Calendly Lecture 17 - hypothesis testing HW 7 out | | | |
| March 28th Lecture 18 - t-tests, errors, experimental bias | Felix OH Calendly HW 7 due @ 11:59pm | Felix OH Calendly 2pm Review (Zoom) | Felix OH Calendly Test 3 | | | HW 7 due |

More recommended resources on these topics

- p-values: YouTube, StatQuest: P values, clearly explained
- p-values: Wikipedia: <https://en.wikipedia.org/wiki/P-value#Calculation>
- Student's t-test, assumptions: https://en.wikipedia.org/wiki/Student%27s_t-test#Assumptions
- Student's t-test (we'll go over this in more depth on Monday): Youtube, Bozeman Science, Student's t-test