

Introduction to Testing

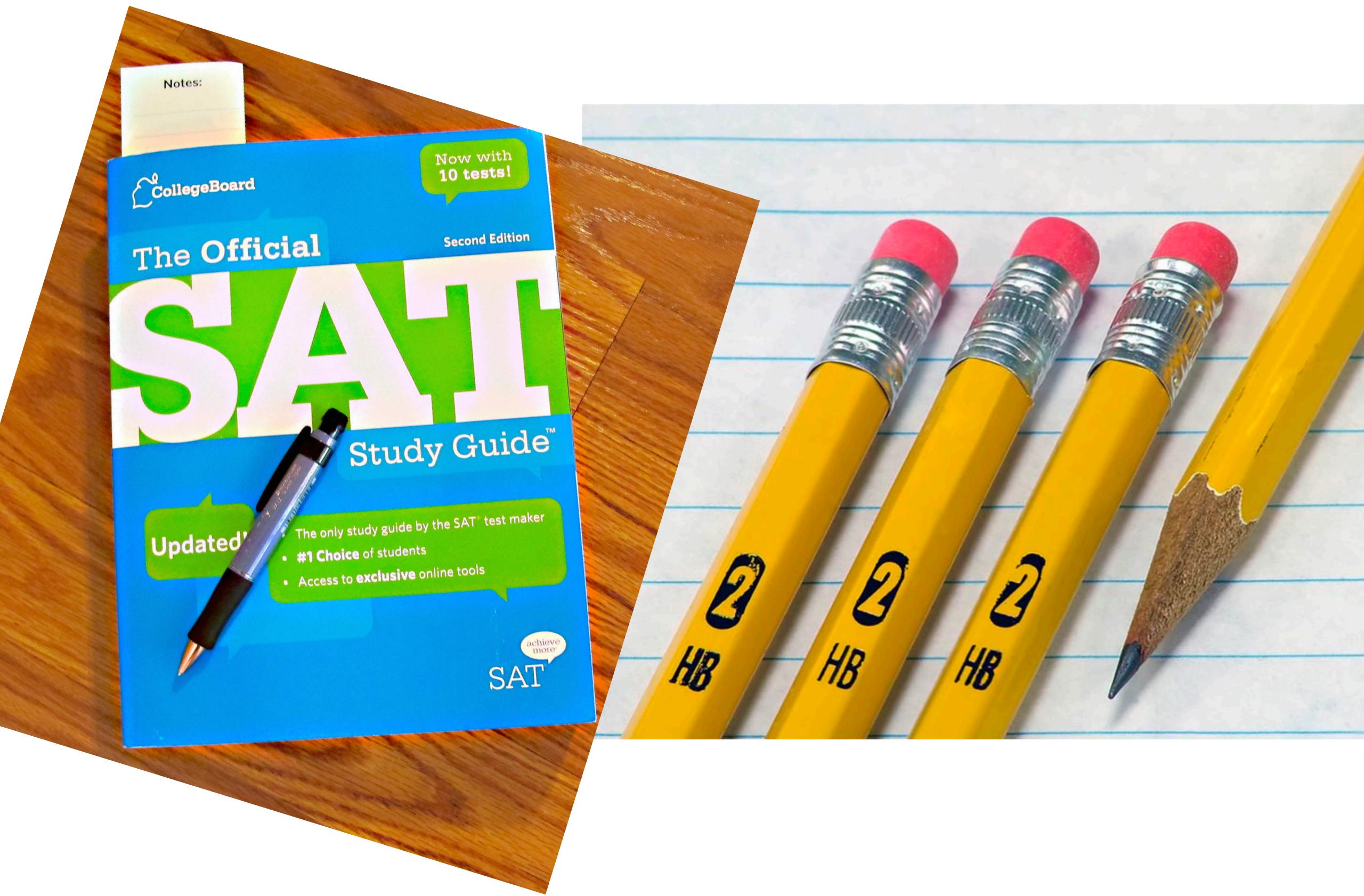
March 15th, 2019

What is testing?

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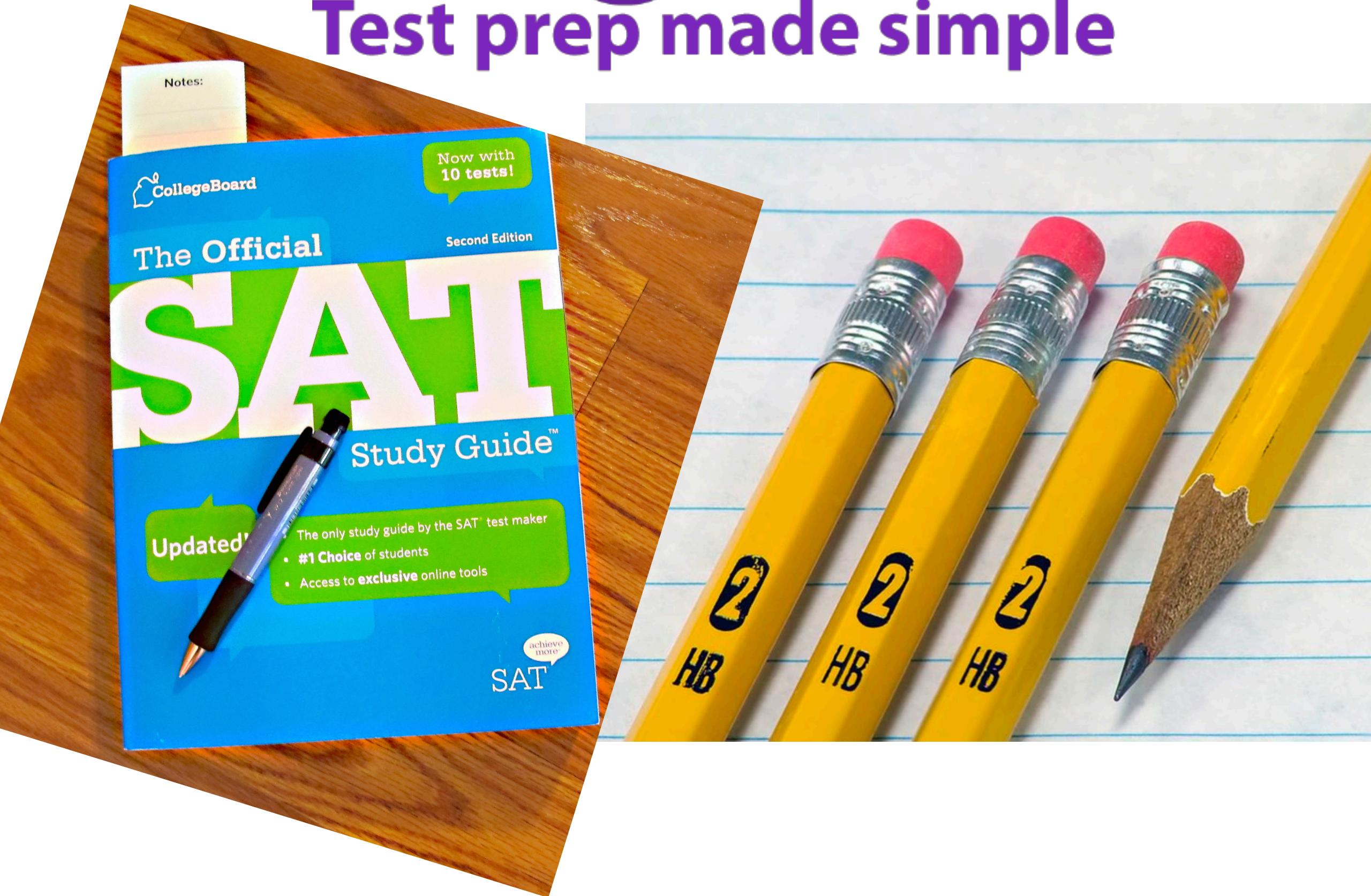


What is testing?



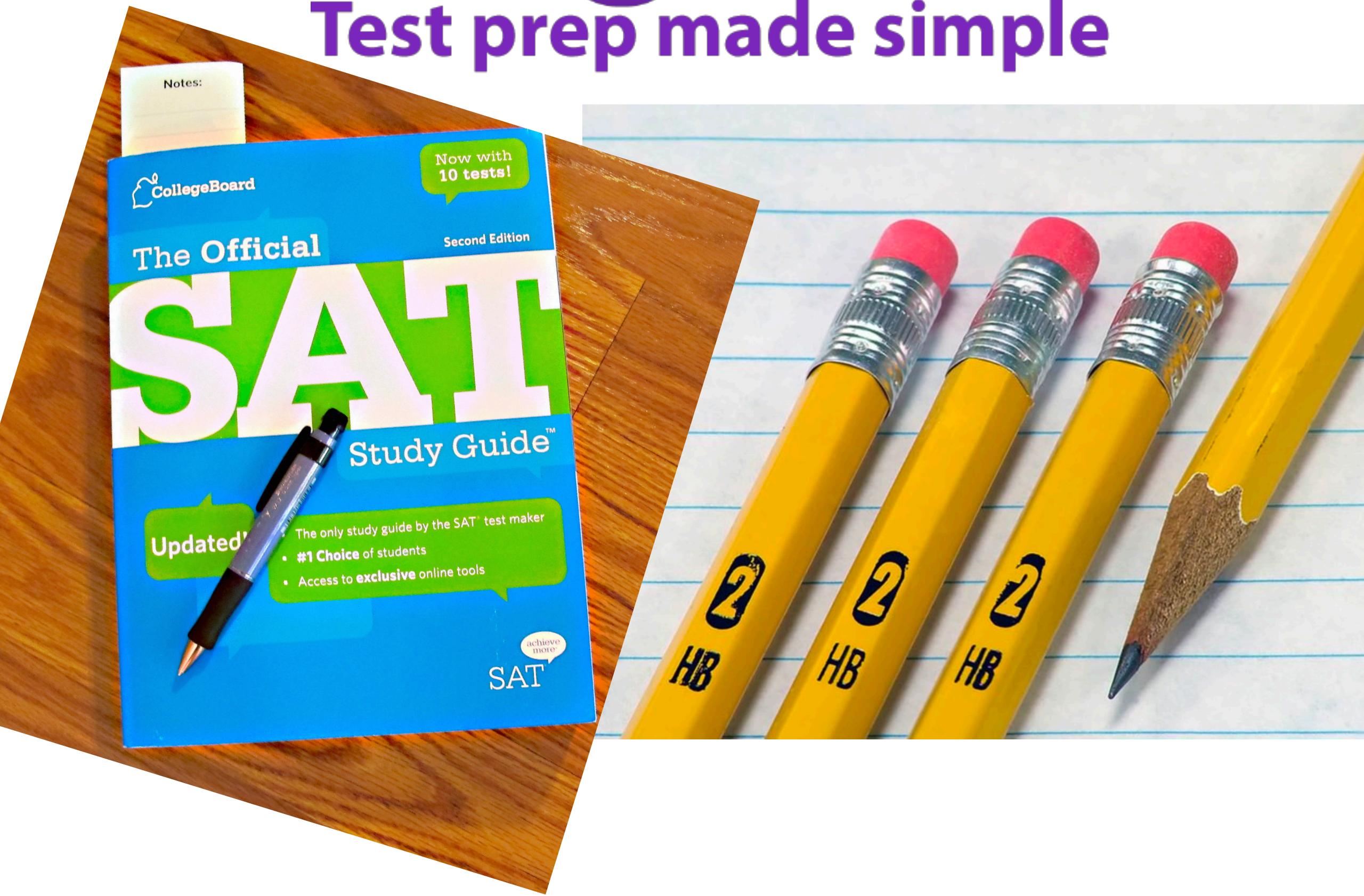
What is testing?

Magoosh
Test prep made simple



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Test prep made simple



What is testing?

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Hypotheses

What is testing?



Hypotheses

Test statistic
or
Confidence
Interval

What is testing?

Hypotheses

Test statistic
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Confidence
Interval

Interpretation
and
Conclusion

Hypotheses

- **Null hypothesis**
 - The status quo
- **Alternative hypothesis**
 - A competing opinion
- **Notation**
 - $H_0 : \mu = 10$

$$H_1 : \mu \neq 10$$

Hypotheses

- **One sided**
 - $H_0 : \mu \geq 9$
 $H_1 : \mu < 9$
- **Two sided**
 - $H_0 : \mu = 10$
 $H_1 : \mu \neq 10$



DISCLAIMER FOR THIS ENTIRE SECTION AND BEYOND

We are not proving anything in hypothesis tests. We are only saying whether or not we reject the null hypothesis based on our data.



**WE CAN ONLY SUGGEST THAT
THE NULL HYPOTHESIS BE
REJECTED**

because our data can tell another story... but as all samples work, not all samples tell the truth. We just understand probabilities associated with finding significant results.



- Three parts
 - Sample estimate (either sample mean or proportion)
 - Critical value $z_{\alpha/2}$ found using `qnorm()`
 - Standard error (see associated sampling distribution)
 - Format

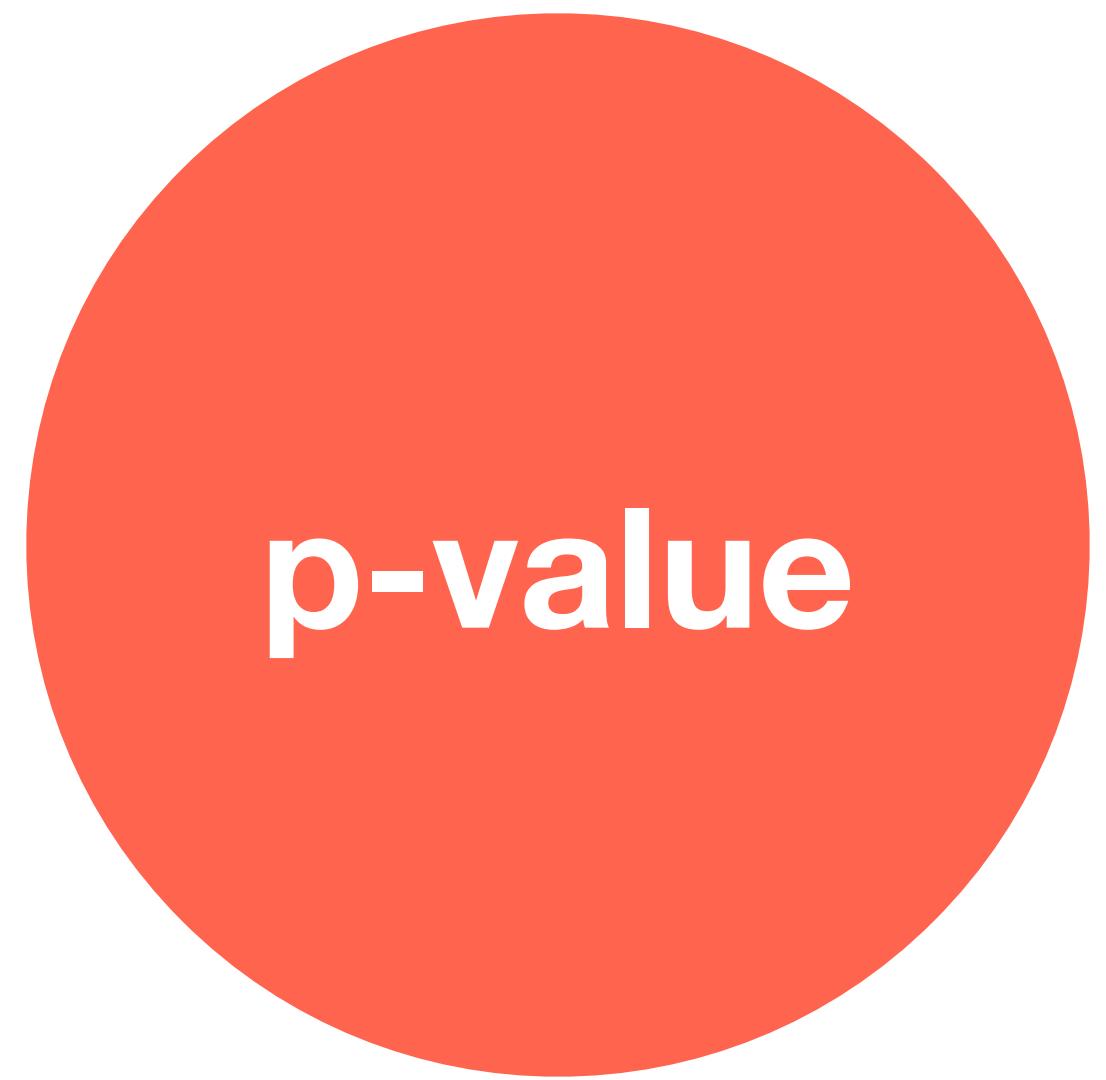
$$\bar{x} \pm (\text{Margin of Error}) = \bar{x} \pm 1.96(\text{SE})$$

For 95% confidence intervals



- **z tests** The most basic of the tests we are covering in this class!
 - 1. Check conditions
 - You have a SRS
 - The underlying population distribution is normal
 - You know the true population standard deviation
 - 2. Make a test statistic
 - 3. Calculate the p-value
 - 4. Interpret p-value
 - 5. Conclude (Will you reject the null?)

$$z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$$



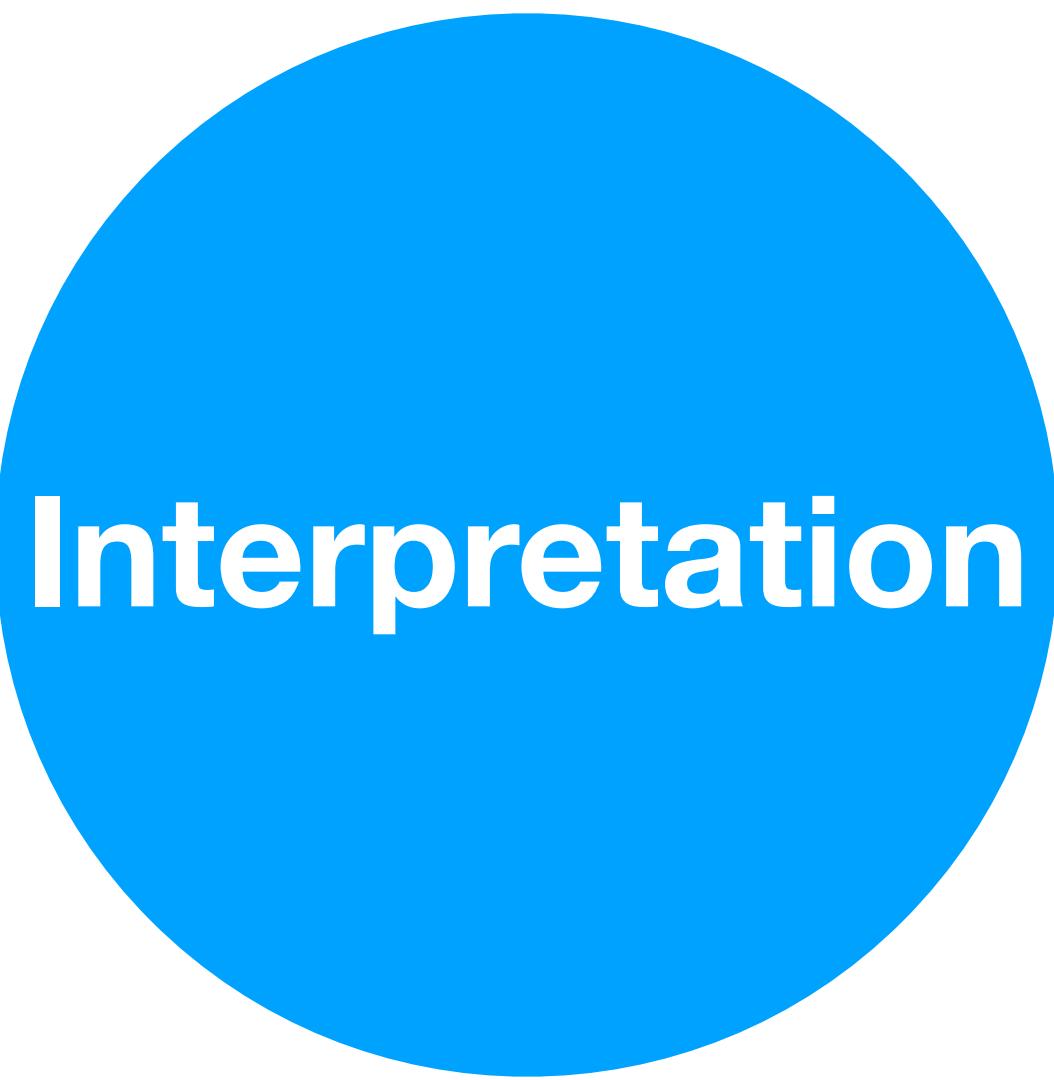
- **p-value**
 - The probability of rejecting the null hypothesis given that the null hypothesis is true
 - The probability of observing our data or more extreme given that the null is true
 - We'll visualize this on the normal distribution for z-tests
 - In general, smaller p-values will imply that we have more evidence against the null hypothesis



Relationship

There is a very specific relationship between confidence intervals and z-tests. For the same data and the same hypotheses, the conclusions of the analyses will be the same.

A 95% confidence interval corresponds to a z test with $\alpha = 0.05$.



- **Confidence Intervals**
 - We are 95% confidence that our true parameter lies within the interval.
 - [Report interval.] This interval was made using a method that creates confidence intervals that contain the true parameter.
- **Test statistic**
 - Our p-value was [this value]. That is, there is a [this value * 100]% chance of observing the data we did or more extreme under the null hypothesis.

Conclusion

- **Confidence Intervals**
 - If our null hypothesized parameter is not within our confidence interval, then we reject the null.
- **Test statistic**
 - If p-value is less than significance level (or very small), then we reject the null hypothesis

Recap... your test will require these things.

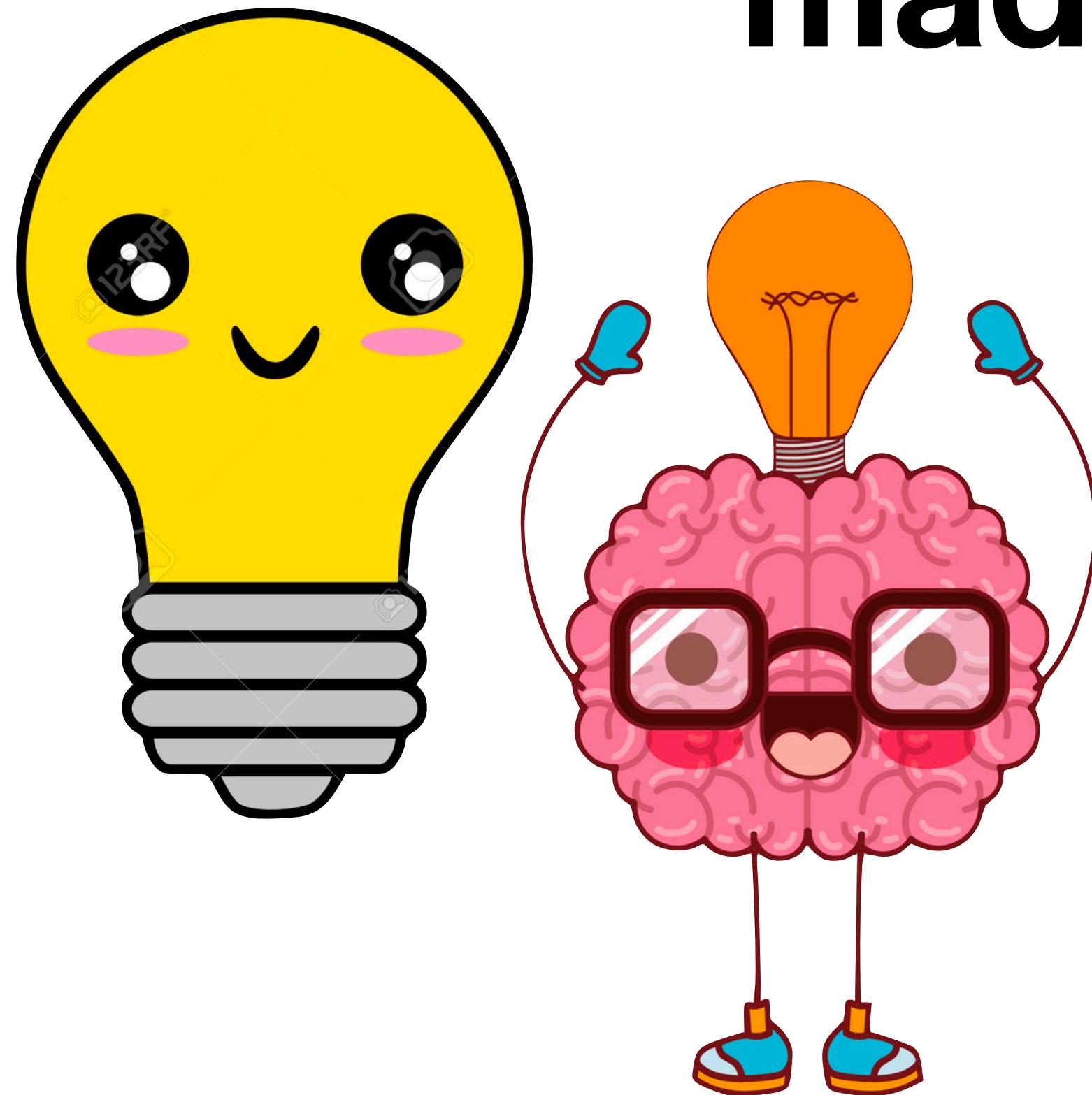
Hypotheses

Test statistic
or
Confidence
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Interpretation
and
Conclusion

Why is this true?

Your hypotheses must be
made before seeing the data.



If you don't, you're treading
into a bad place.

If you don't, you're treading
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into a bad place.

Meet this guy.

**Meet this guy.
He's a mad scientist.**

**Meet this guy.
He's a mad scientist.
He will do anything to prove a point.**

**Meet this guy.
He's a mad scientist.
He will do anything to prove a point.**



**Joobika , Jumba
300 lbs
Current location: Hawaii**

The golden rule in research is that if you have a p-value of less than 0.05, then you found a significant discovery.

By chance alone,
we may be able to get a small p-value
based on your sample.

People who abuse the above
are called “p-hackers”.

**The golden rule in research is that if you have a
p-value of less than 0.05, then you found a
significant discovery.**

**By chance alone,
we may be able to get a small p-value
based on your sample. If you just keep running
your experiment a million times, then at least one
of your tests can be significant.**

And by dishonesty, **you can fudge your data.**

**People who abuse the above
are called “p-hackers”.**

**Those who abuse the science of p-values
are called “p-hackers”.**

**But if you are a RESEARCHER
with a DOCTORATE, why
would you do this?!**

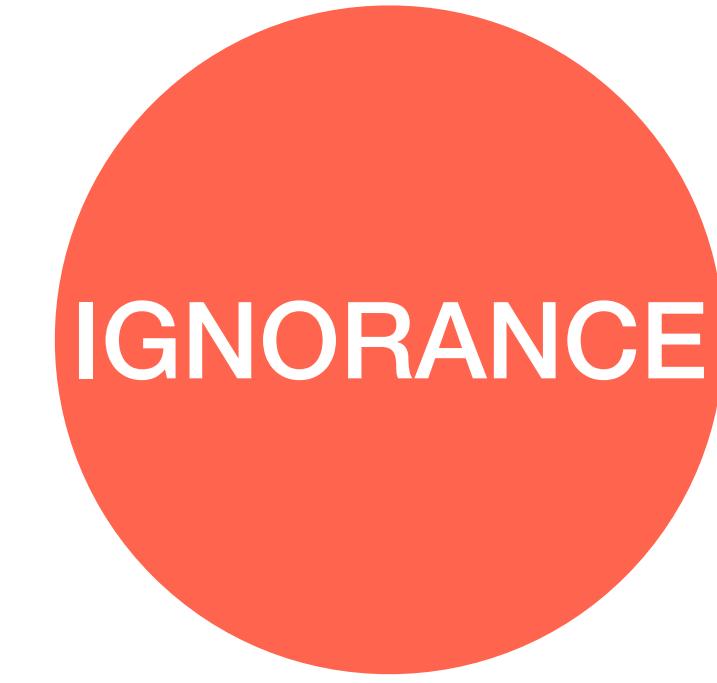


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BLIND
PASSION

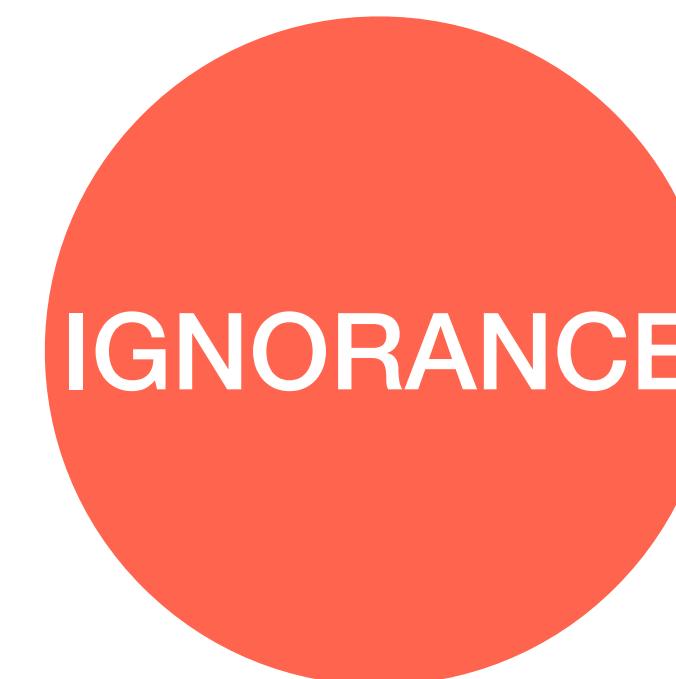
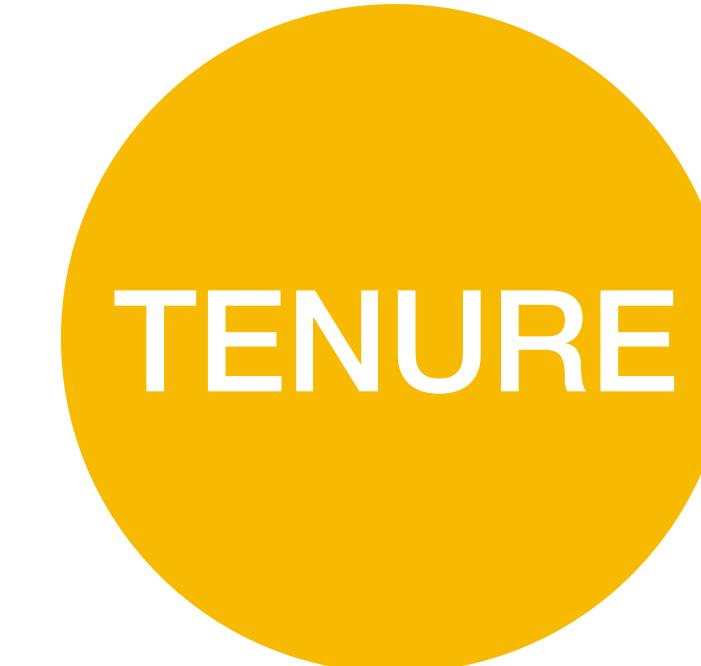
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IGNORANCE

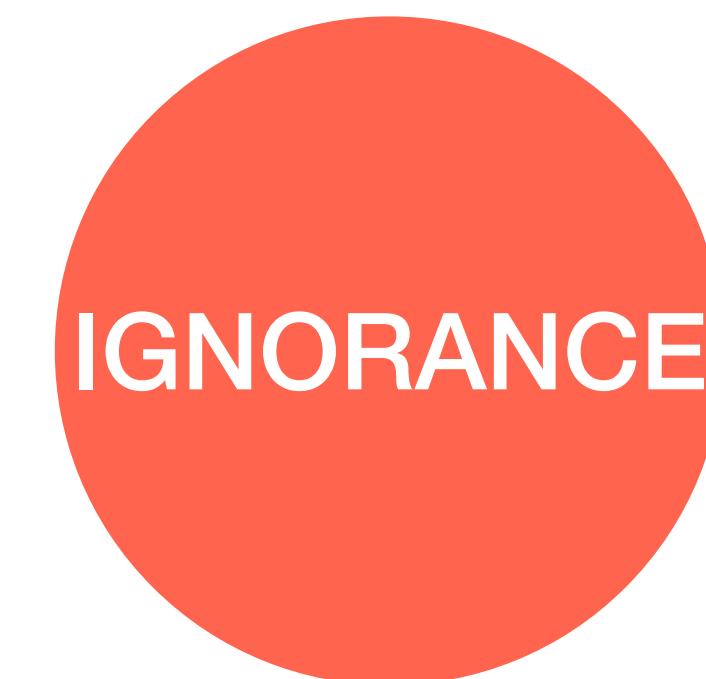
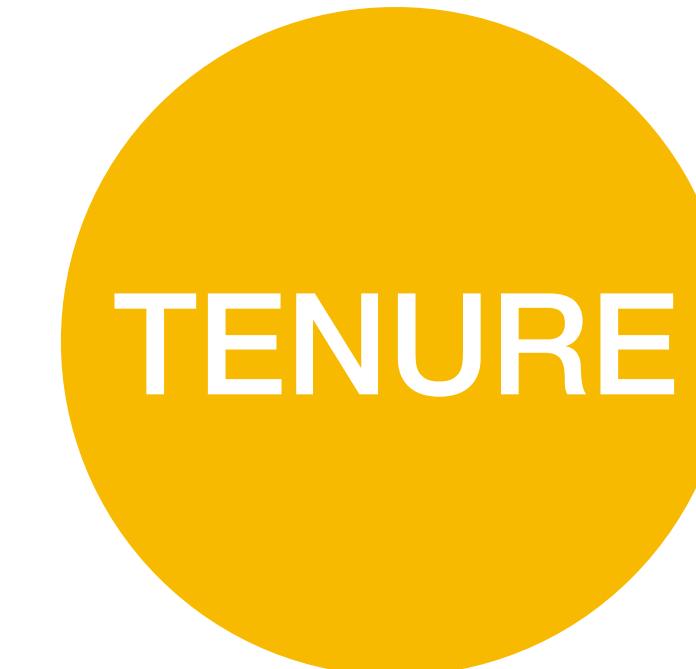


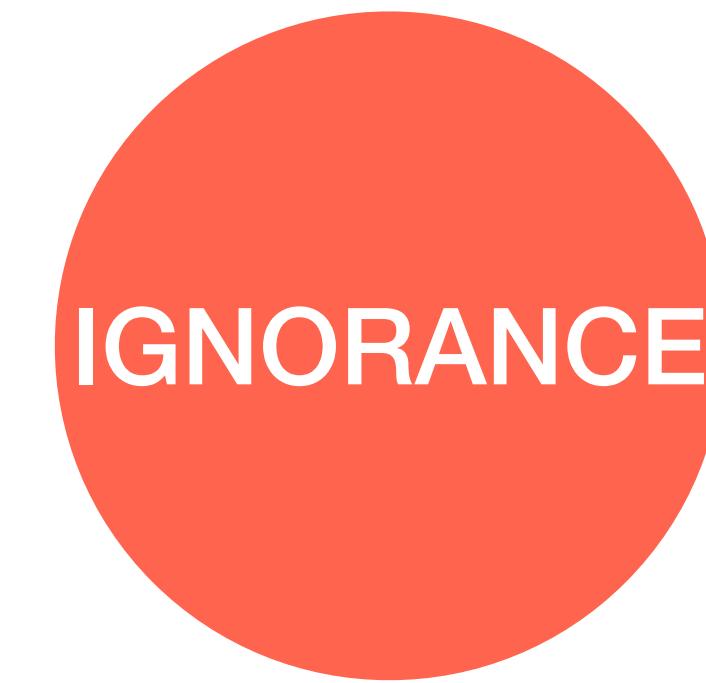
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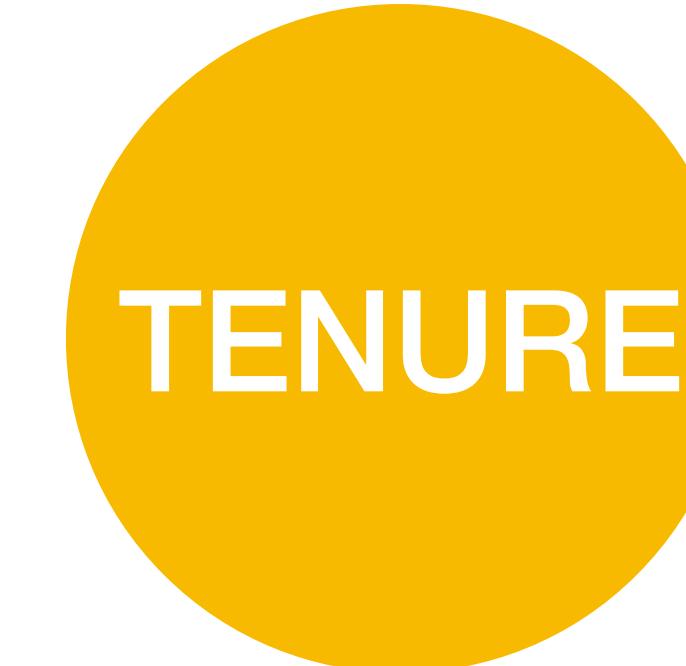


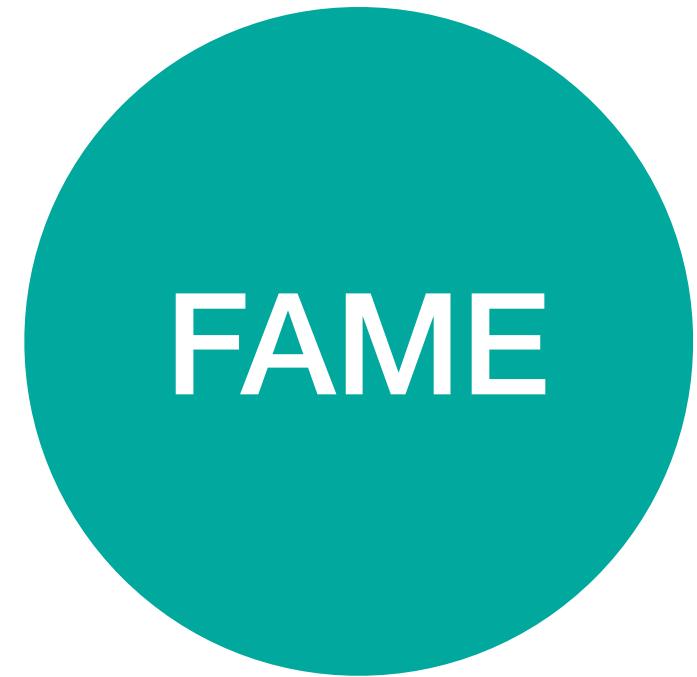
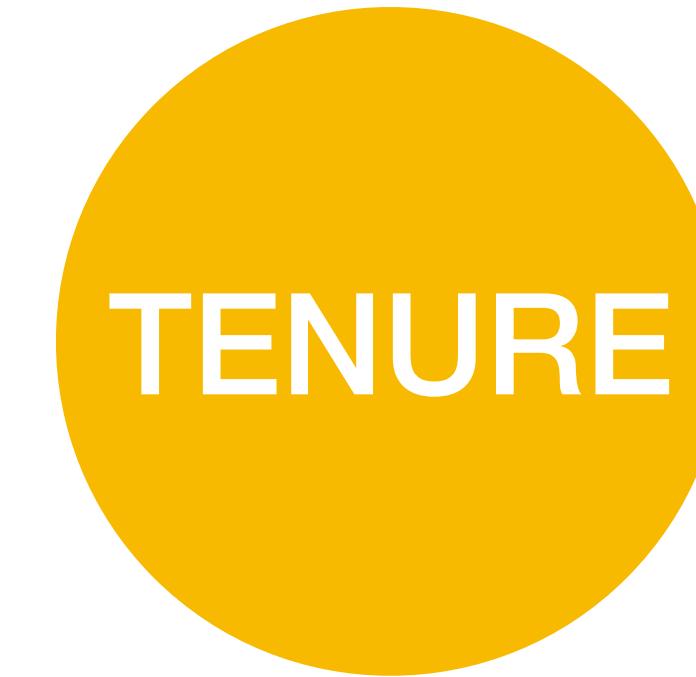
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Top Cornell food researcher **Brian Wansink** did it.



Review

Binomial

You can approximate binomial as normal based on certain conditions.

When n is large (and $np > 10$, $n(1-p) > 10$), then $\text{Bin}(n,p)$ is approximately $N(np, \sqrt{np(1-p)})$.

$\text{mean}(\text{binomial}) = np$
 $\text{sd}(\text{binomial}) = \sqrt{np(1-p)}$

Both are discrete distributions

Poisson

This is continuous.

Normal