

Hypothesis Tests and Confidence Intervals using Normal Distribution!

STAT 120

Sections 5.1-5.2

Day 15

A Research Question

How do Malaria parasites impact mosquito behavior?



Malaria Parasites and Mosquitoes

- Mosquitos were randomized to either eat from a malaria infected mouse (exposed group) or a healthy mouse (control group)
- After infection, the parasites go through two stages:
 - 1) Not yet infectious (Days 1-8)
 - 2) Infectious (Days 9 – 28)
- Response variable: whether the mosquito approached a human (in a cage with them)
- Does this behavior differ by exposed vs control? Does it differ by infection stage?

Cator LJ, George J, Blanford S, Murdock CC, Baker TC, Read AF, Thomas MB. (2013). ['Manipulation' without the parasite: altered feeding behaviour of mosquitoes is not dependent on infection with malaria parasites.](#) Proc R Soc B 280: 20130711.

Malaria Parasites and Mosquitoes

- Does infecting mosquitoes with Malaria actually impact the mosquitoes' behavior to favor the parasite?
- Malaria parasites would benefit if:
 - ▣ Mosquitoes approached humans less often after being exposed, but before becoming infectious, because humans are risky
 - ▣ Mosquitoes approached humans more often after becoming infectious, to pass on the infection

We'll first look at the mosquitoes **before they become infectious** (days 1–8).

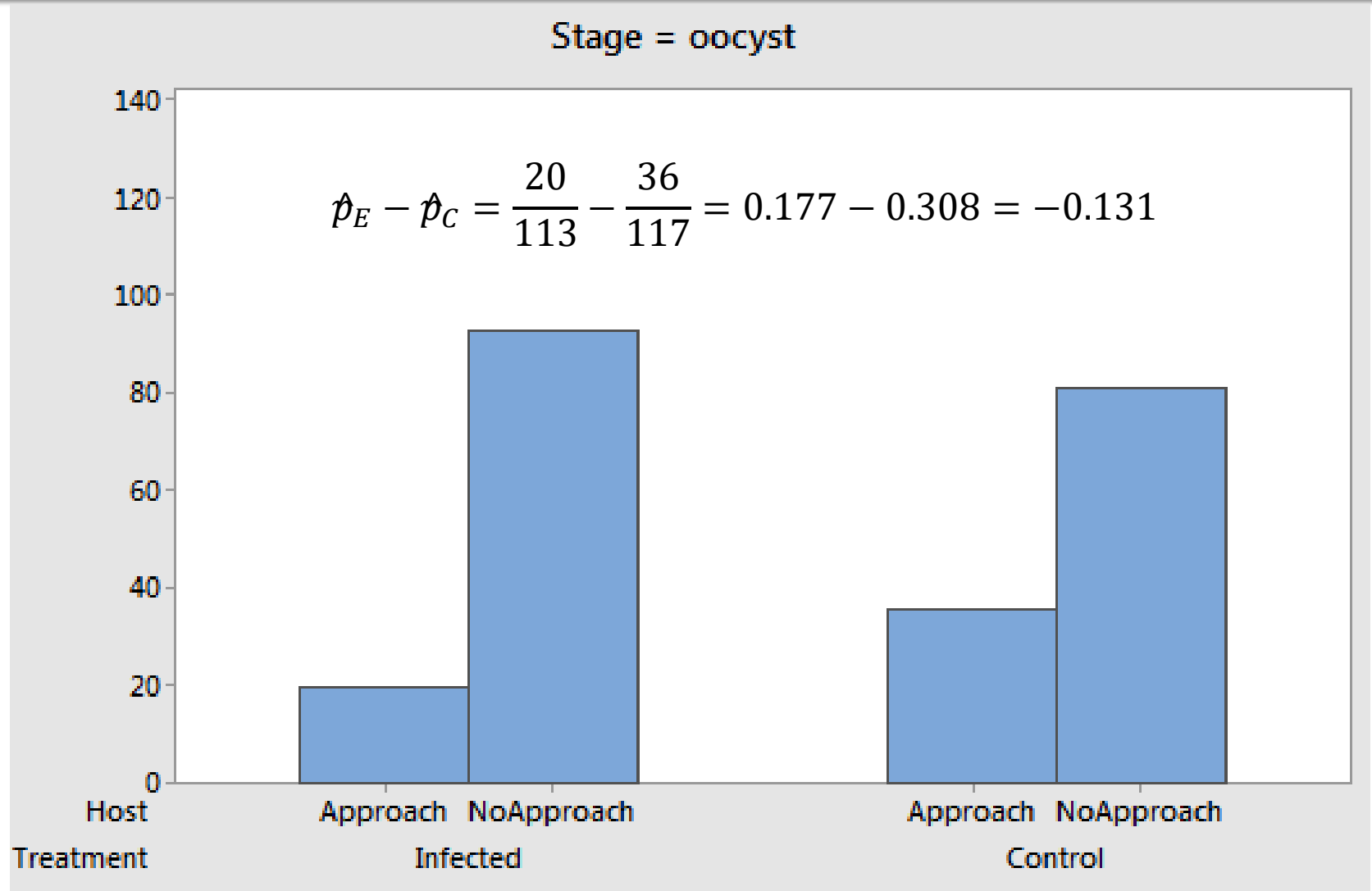
p_C : proportion of controls to approach human

p_E : proportion of exposed to approach human

What are the relevant hypotheses?

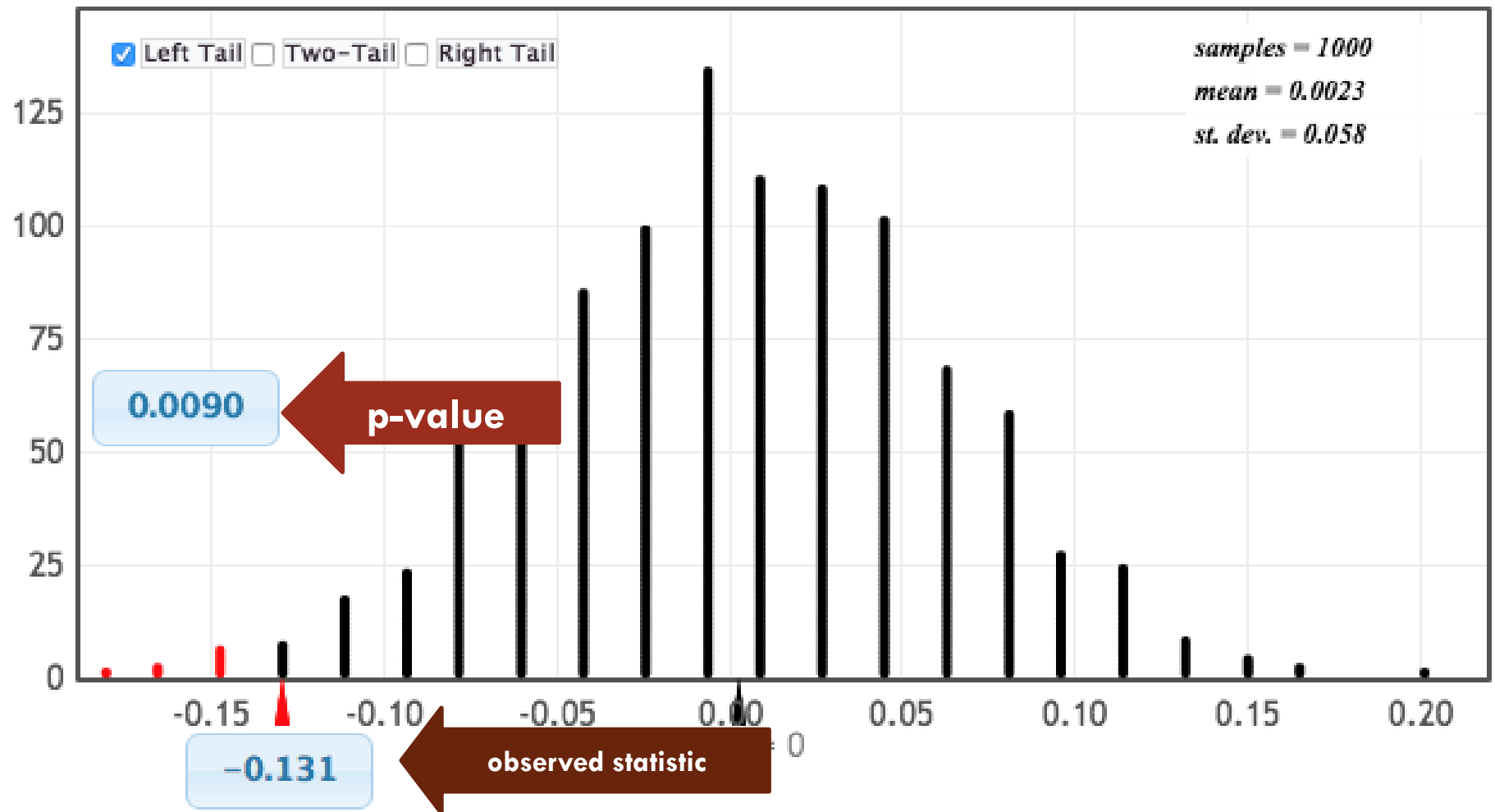
- A. $H_0: p_E = p_C, H_a: p_E < p_C$
- B. $H_0: p_E = p_C, H_a: p_E > p_C$
- C. $H_0: p_E < p_C, H_a: p_E = p_C$
- D. $H_0: p_E > p_C, H_a: p_E = p_C$

Data: Before Infectious



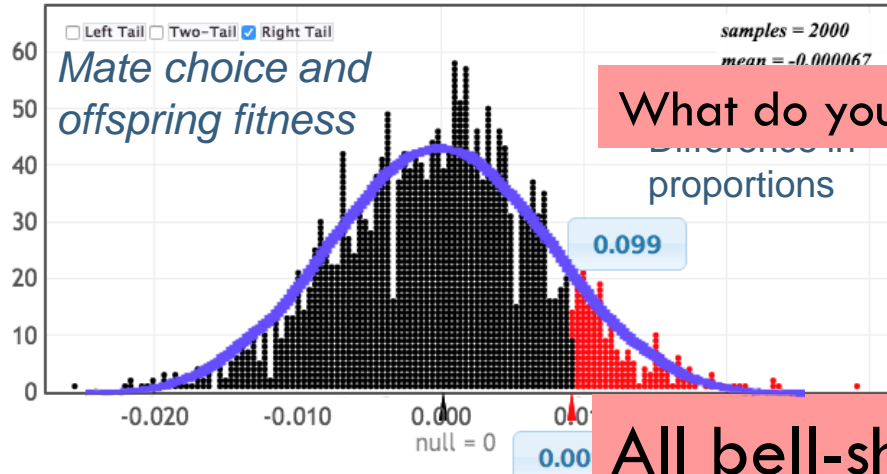
Randomization Test

Randomization Dotplot of $\hat{p}_1 - \hat{p}_2$ Null Hypothesis: $p_1 = p_2$

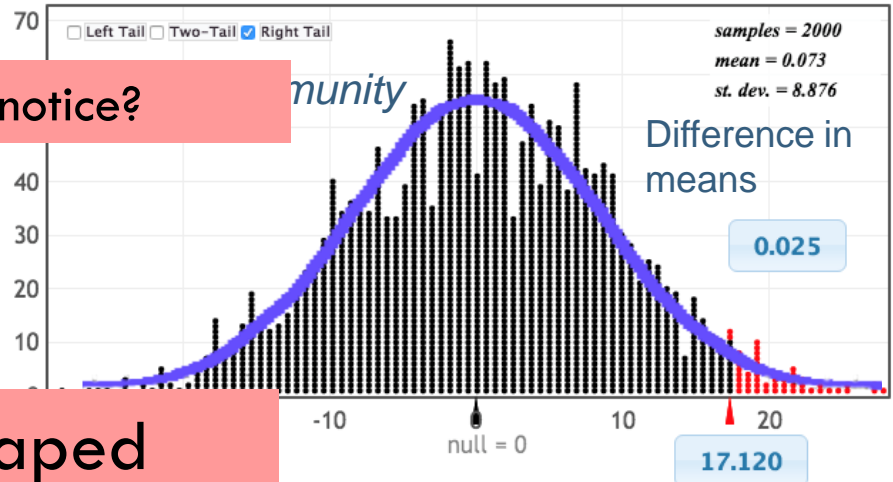


Randomization and Bootstrap Distributions

Randomization Dotplot of $\hat{p}_1 - \hat{p}_2$ Null Hypothesis: $p_1 = p_2$



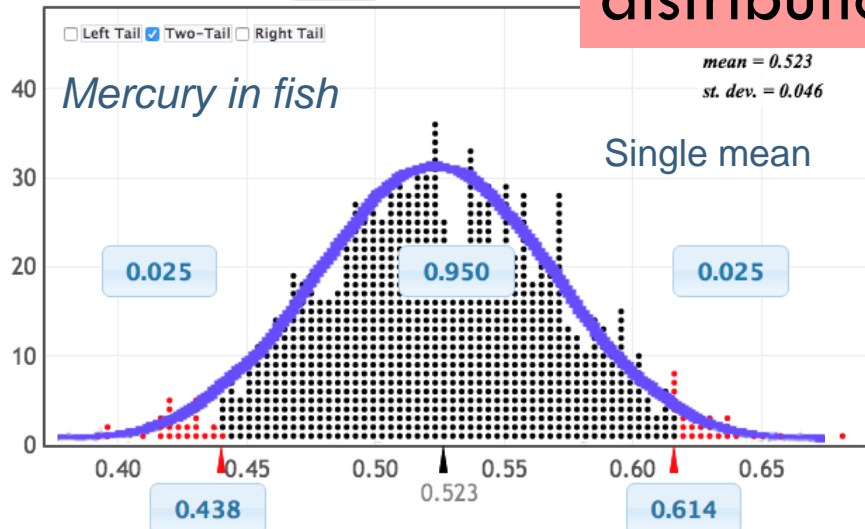
Randomization Dotplot of $\bar{x}_1 - \bar{x}_2$, Null hypothesis: $\mu_1 = \mu_2$



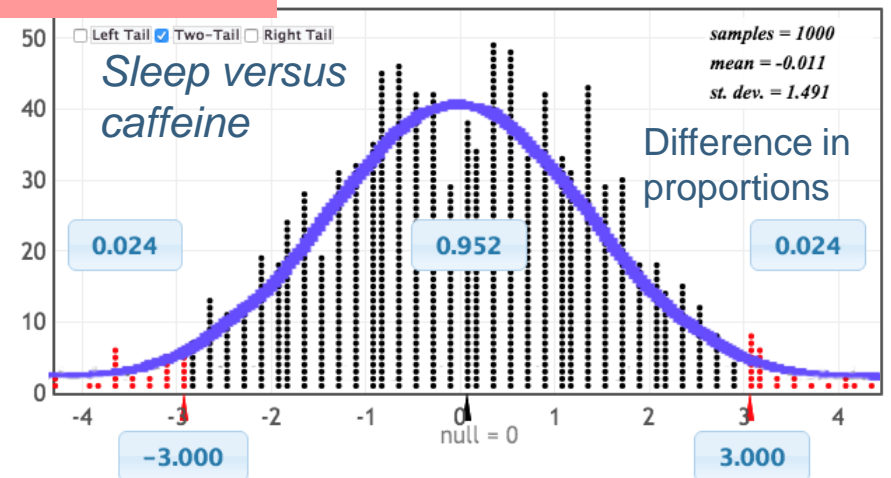
What do you notice?

All bell-shaped distributions!

Bootstrap Dotplot of Mean



Randomization Dotplot of $\bar{x}_1 - \bar{x}_2$, Null hypothesis: $\mu_1 = \mu_2$



Central Limit Theorem

For random samples with a sufficiently large sample size, the distribution of sample statistics for a **mean** or a **proportion** is **normally distributed**

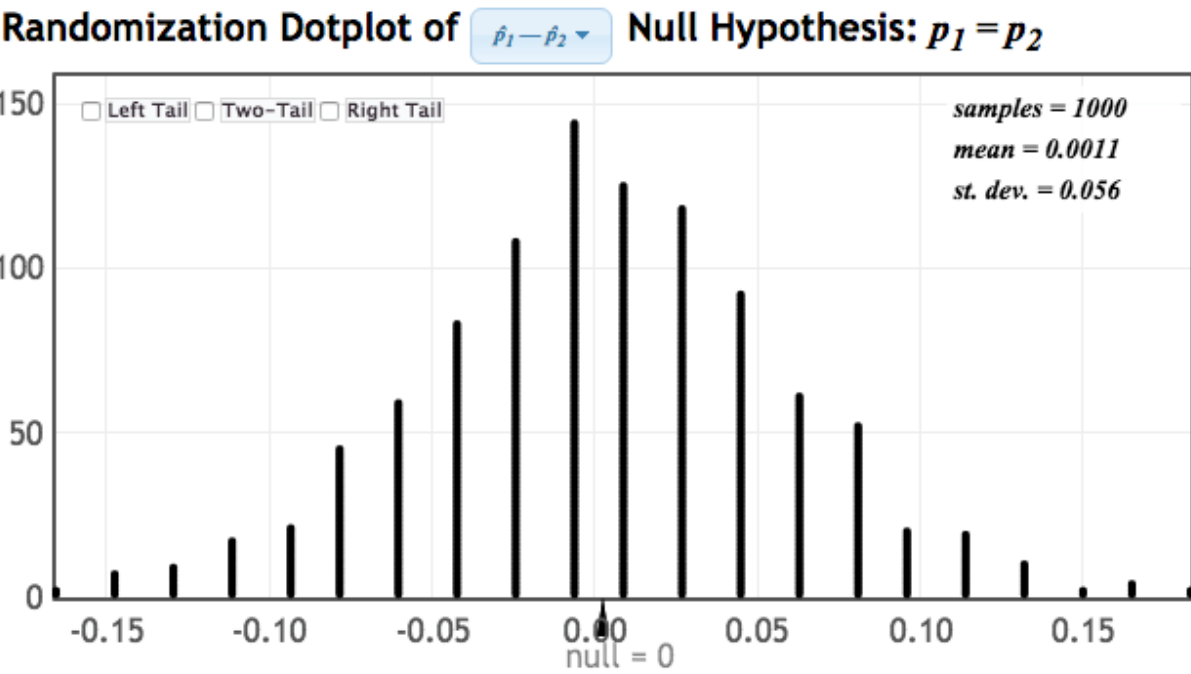
Central Limit Theorem

- The catch:

“sufficiently large sample size”

- The **more skewed** the original distribution of data/population is, the larger n has to be for the CLT to work
- For quantitative variables that are not very skewed, $n \geq 30$ is usually sufficient
- For categorical variables, counts of at least 10 within each category is usually sufficient

Which normal distribution should we use to approximate this?



Original Sample

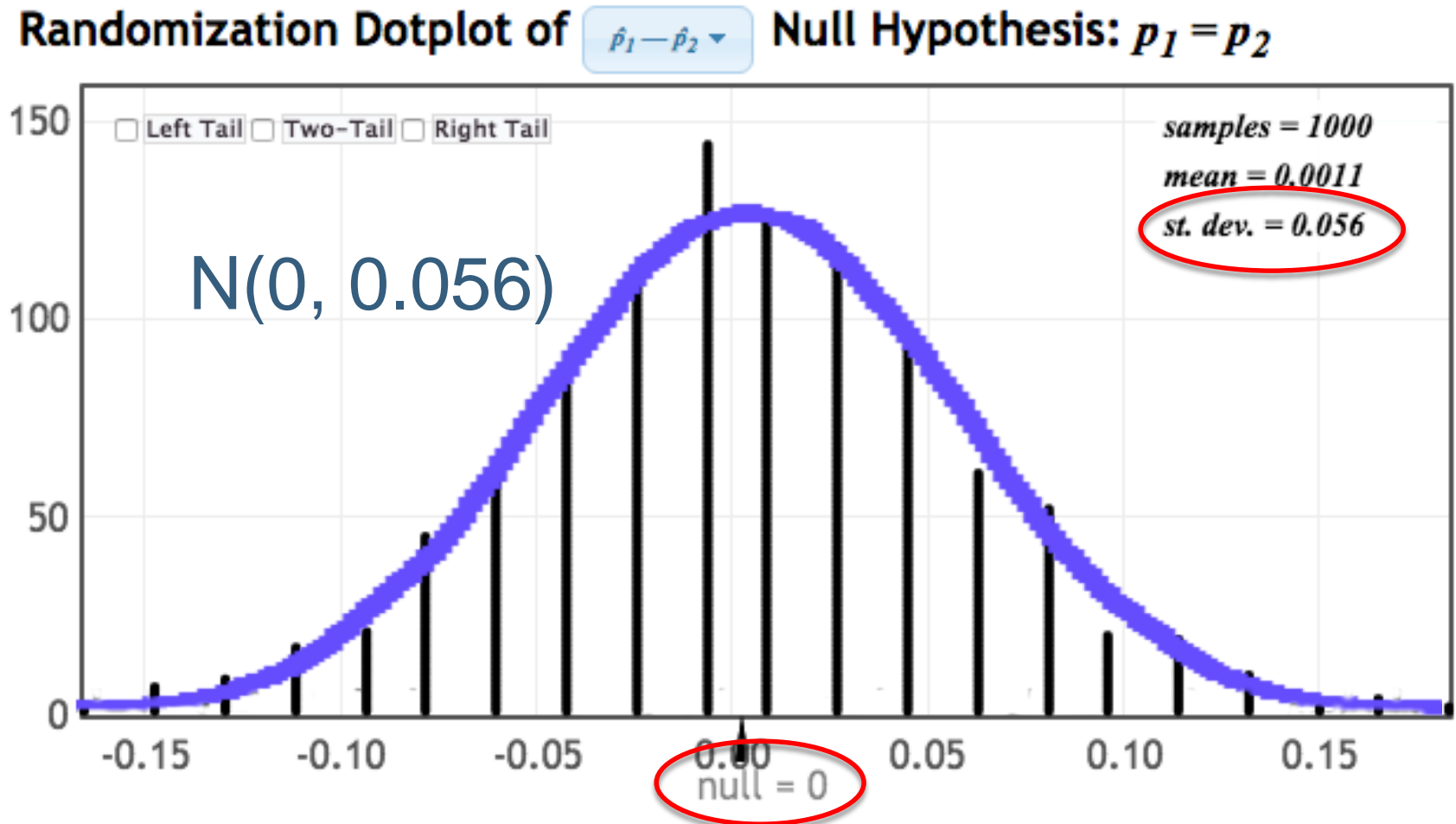
Group	Count	Sample Size	Proportion
Group 1	20	113	0.177
Group 2	36	117	0.308
Group 1-Group 2	-16	n/a	-0.131

Randomization Sample

Group	Count	Sample Size	Proportion
Group 1	27	113	0.239
Group 2	29	117	0.248
Group 1-Group 2	-2	n/a	-0.0089

- A. $N(0, -0.131)$
- B. $N(0, 0.056)$
- C. $N(-0.131, 0.056)$
- D. $N(0.056, 0)$

Normal Distribution



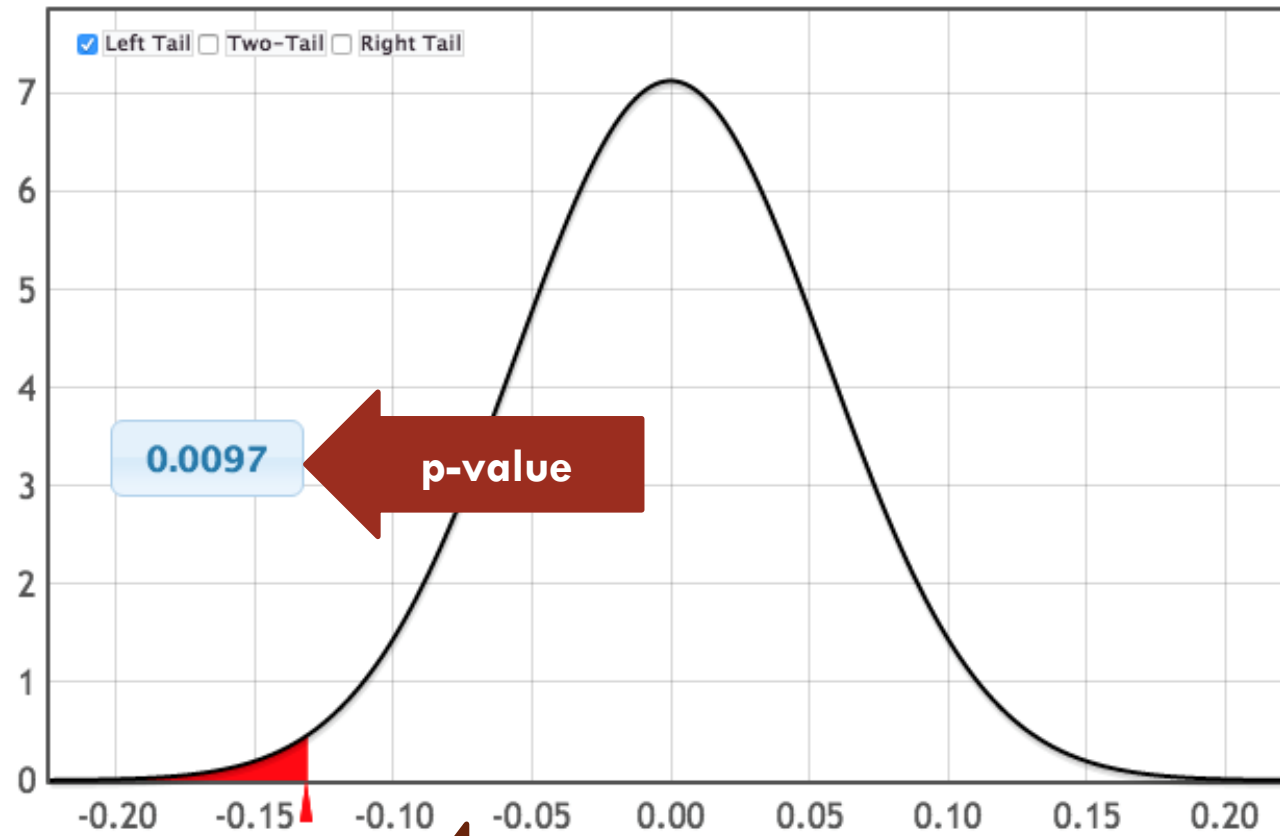
We can compare the original statistic to this Normal distribution to find the p-value!

p-value from $N(\text{null}, \text{SE})$

Normal Distribution ▾

Reset Plot

☒ Left Tail ☐ Two-Tail ☐ Right Tail



Normal Distribution

Mean

Standard Deviation

0

0.056

Edit Parameters

Connecting Normal model to hypothesis tests

- Hypothesis test:
 - Suppose: randomization distribution is bell-shaped.
 - Center: hypothesized null parameter value
 - Spread: the standard error given in the randomization graph (or by formula)
- P-value is computed from the normal model the “usual” way – the chance of being as extreme, or more extreme, than the observed statistic.

Standardized Statistic

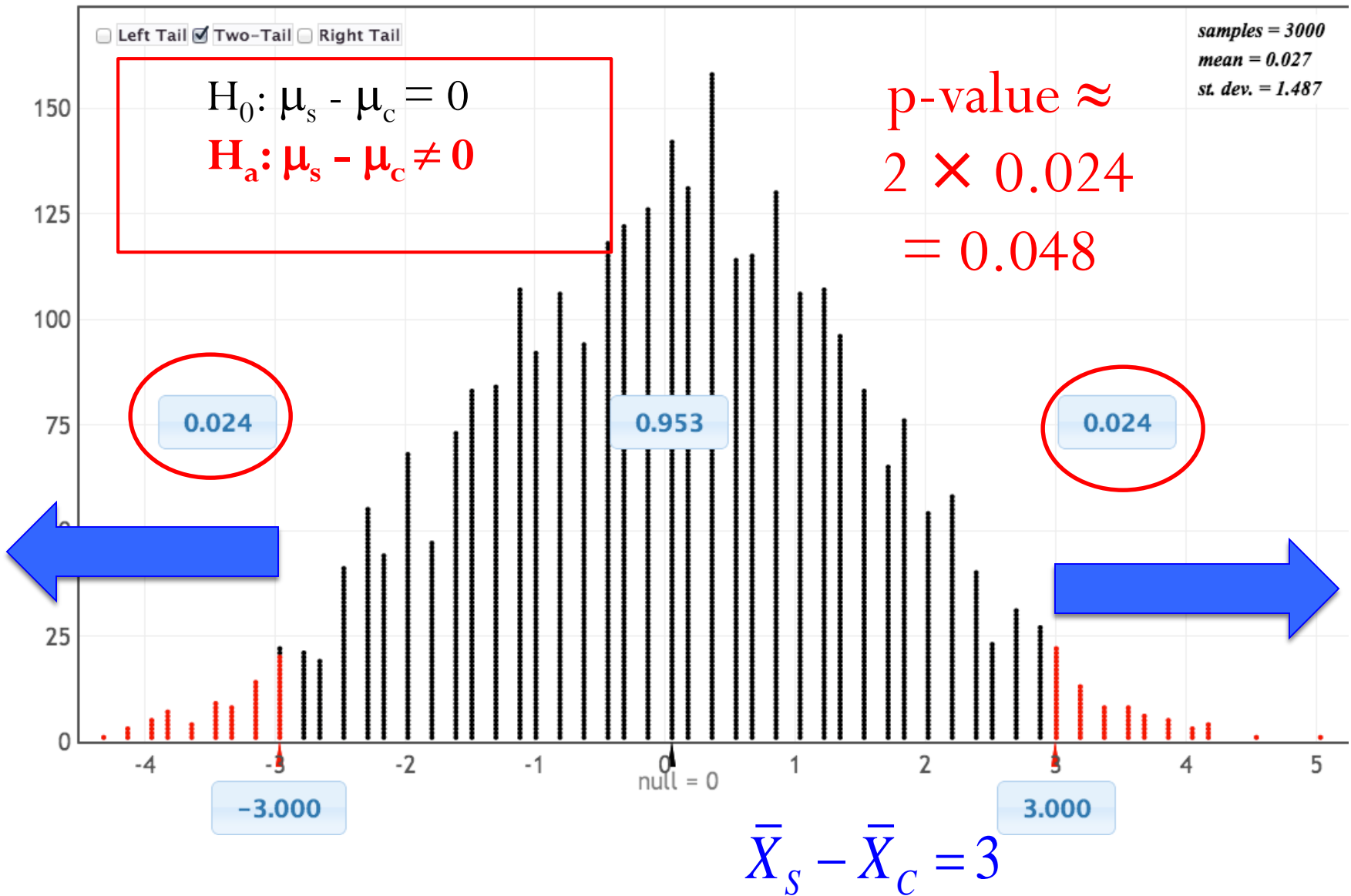
The **standardized test statistic** (also known as a z-statistic) is

$$z = \frac{\textit{statistic} - \textit{null}}{SE}$$

- Calculating the number of standard errors a statistic is from the null lets us assess extremity on a common scale.

Sleep versus Caffeine

Randomization Dotplot of $\bar{x}_I - \bar{x}_2$, Null hypothesis: $\mu_I = \mu_2$



Sleep versus Caffeine



$$H_0: \mu_s - \mu_c = 0$$

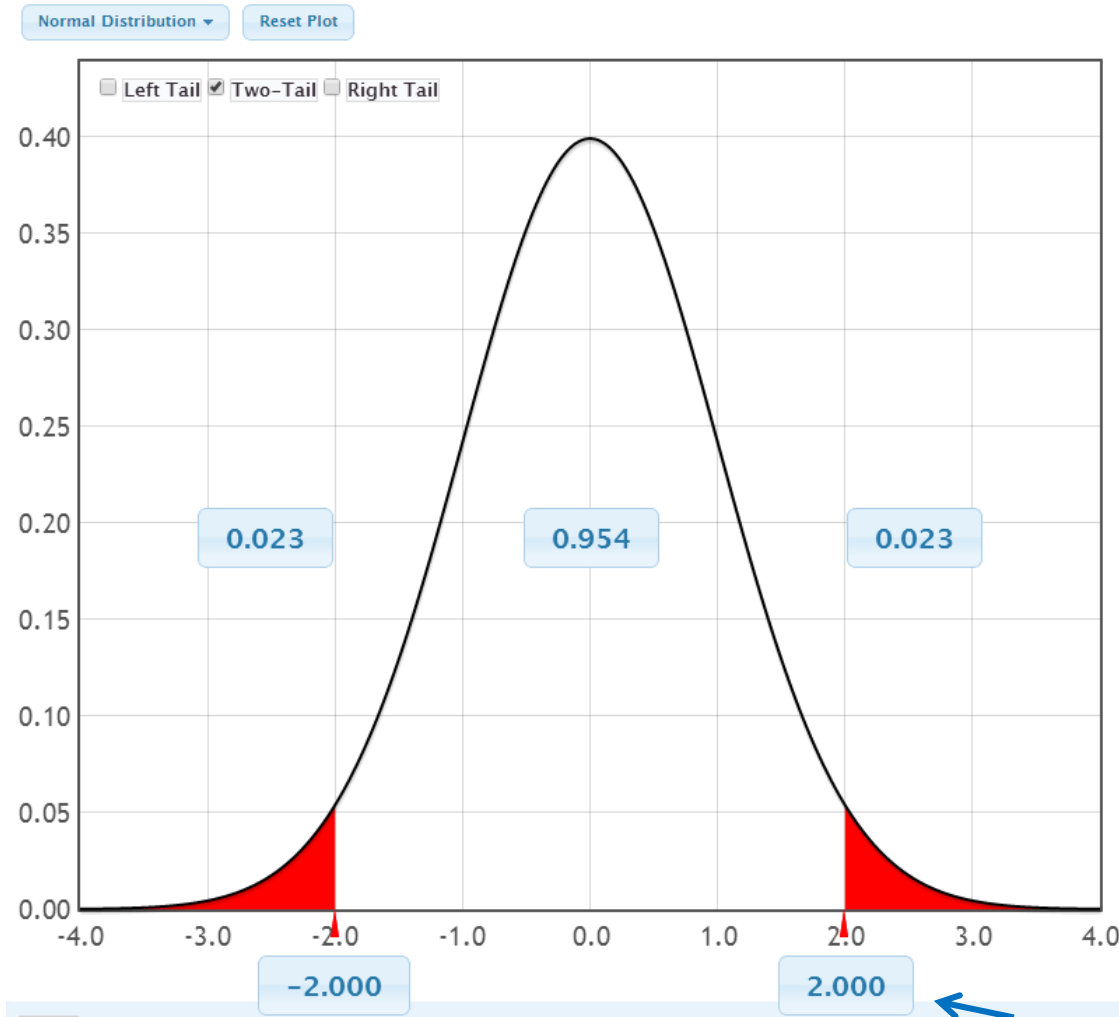
$$H_a: \mu_s - \mu_c \neq 0$$

- The observed difference is 3 words.
- The randomization model is approximately $N(0, 1.5)$.
- The observed difference is 2 SEs above the null difference of 0:

$$z = \frac{\text{sample stat} - \text{null parameter value}}{\text{SE}} = \frac{3 - 0}{1.5} = 2$$

- The p-value is the area above +3 and below -3 on the $N(0, 1.5)$ model.
- Equivalently, the p-value is the chance of getting a mean difference more than 2 SEs away from the hypothesized mean difference of 0.

Sleep versus Caffeine



Normal Distribution

Mean	Standard Deviation
0	1

$N(0,1)$ model!

The p-value is about $2(.023)$, or about 4.6%.

```
> 2*pnorm(-2)  
[1] 0.04550026
```

Standardized Test Stat value

Malaria and Mosquitos

- Does infecting mosquitoes with Malaria actually impact the mosquitoes' behavior to favor the parasite?
- **After the parasite becomes infectious**, do infected mosquitoes approach humans **more** often, so as to pass on the infection?

For the data **after the mosquitoes become infectious (Days 9 – 28)**, what are the relevant hypotheses?

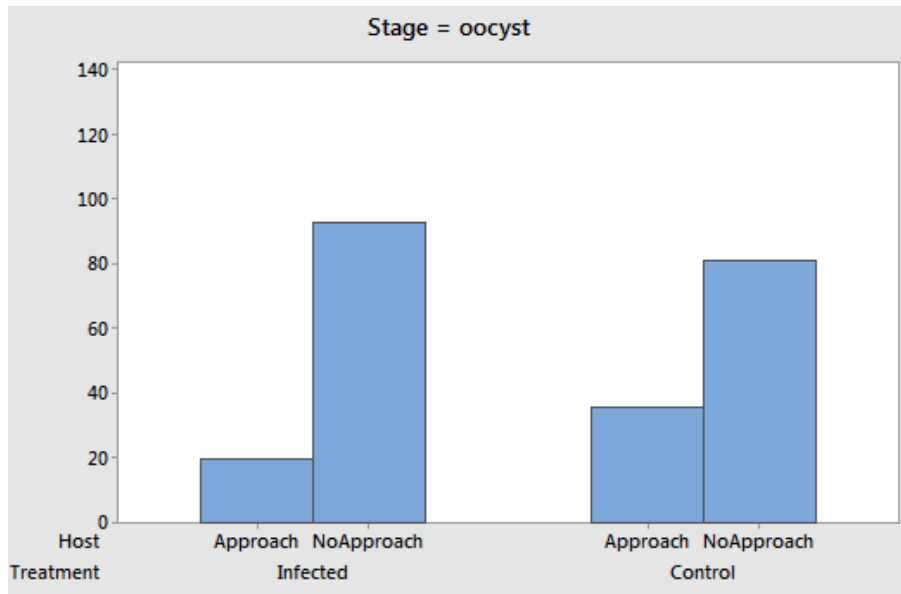
p_C : proportion of controls to approach human

p_E : proportion of exposed to approach human

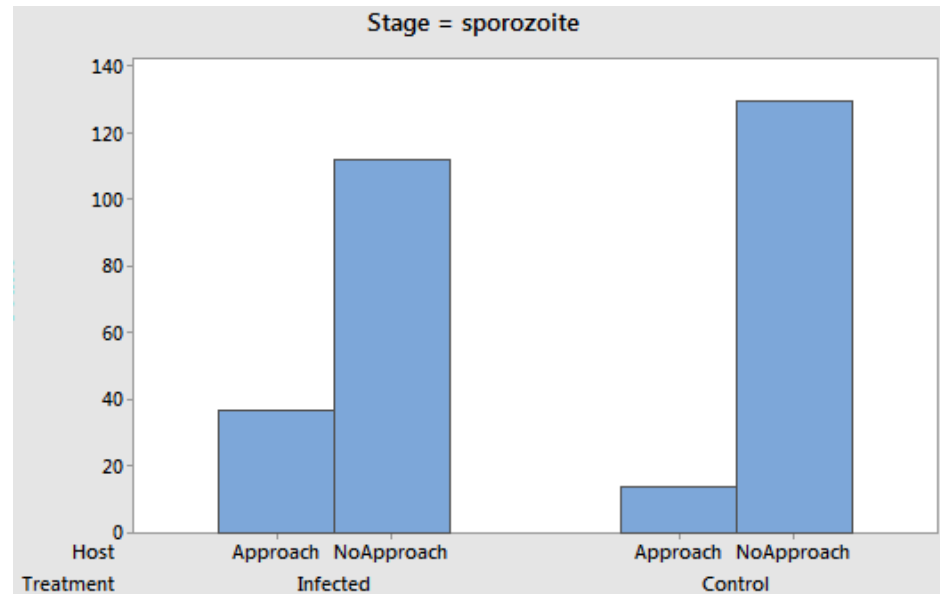
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- D. $H_0: p_E > p_C, H_a: p_E = p_C$

Data

Before Infectious

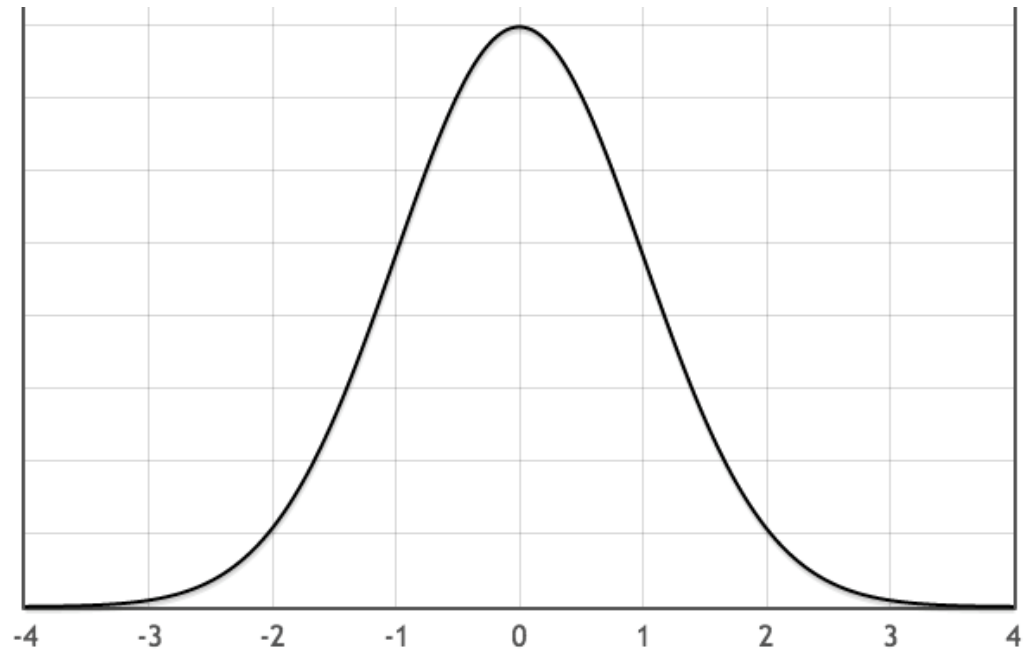


After Infectious



The difference in proportions is 0.151 and the standard error is 0.05. Is this significant?

- A. Yes
- B. No



Malaria and Mosquitoes

- It appears that mosquitoes infected by malaria parasites do, in fact, behave in ways advantageous to the parasites!
 - ▣ Exposed mosquitos are **less** likely to approach before becoming infectious (so more likely to stay alive)
 - ▣ Exposed mosquitos are **more** likely to approach humans after becoming infectious (so more likely to pass on disease)
- Toxoplasmosis parasites impact rat and human behavior, malaria parasites impact mosquito behavior... what else might parasites be impacting that we have yet to discover??

Formula for p-values Using $N(0,1)$

From original data

From H_0

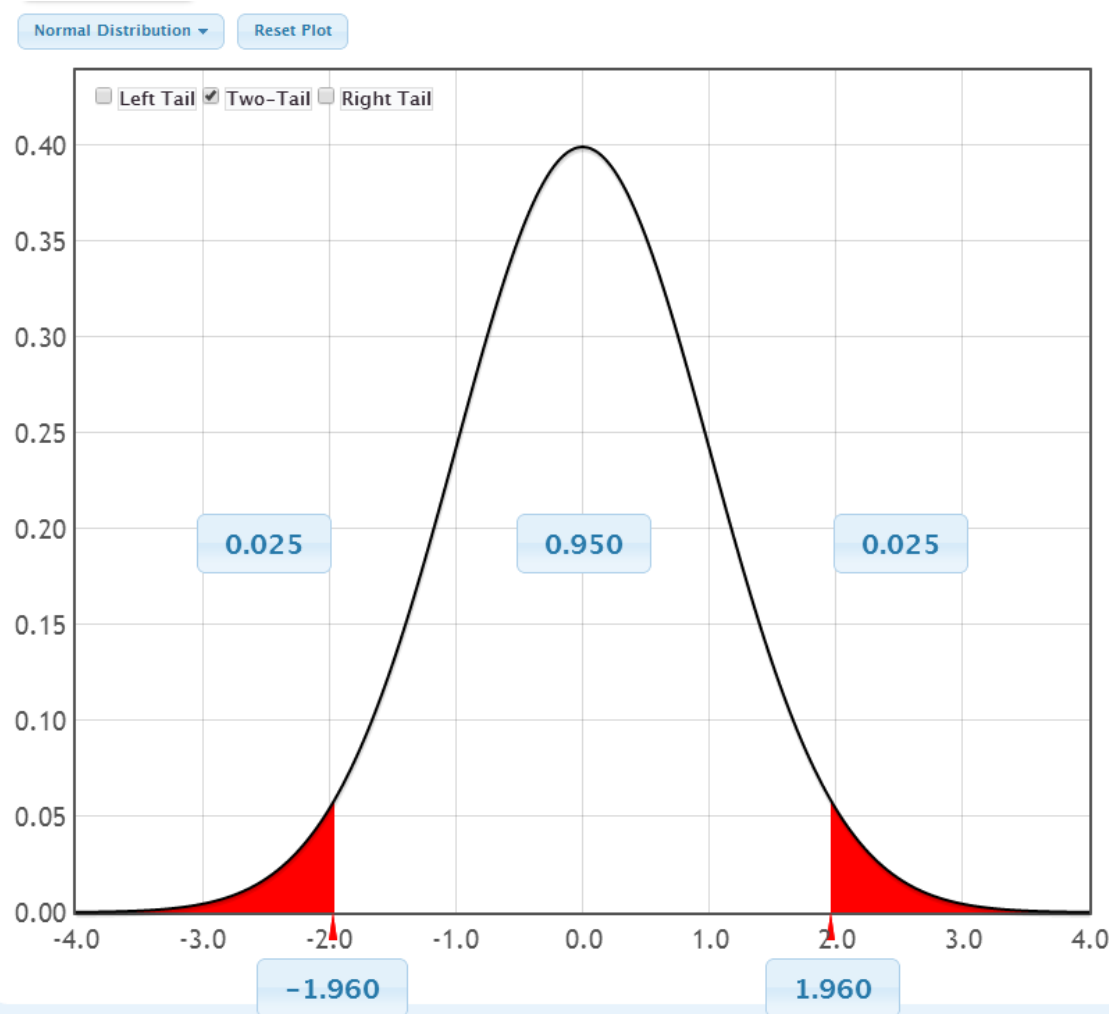
$$z = \frac{\text{sample statistic} - \text{null value}}{\text{SE}}$$

From randomization distribution

Connecting Normal model to Confidence Intervals

- Confidence Intervals
 - Suppose: bootstrap distribution is bell-shaped.
 - Center: sample statistic
 - Spread: the standard error given in the bootstrap graph (or by formula)
- To get a 95% confidence interval we compute
$$\text{statistic} \pm 2(\text{SE})$$
- Why 2 SE's?
 - 95% of all sample means fall within 2 SE's of the population mean*
 - The value 2 is a z-score!
 - * well, actually the precise z-score under a normal model is $z=1.96$ instead of 2!

N(0,1) model



Normal Distribution

Mean	Standard Deviation
0	1

Edit Parameters

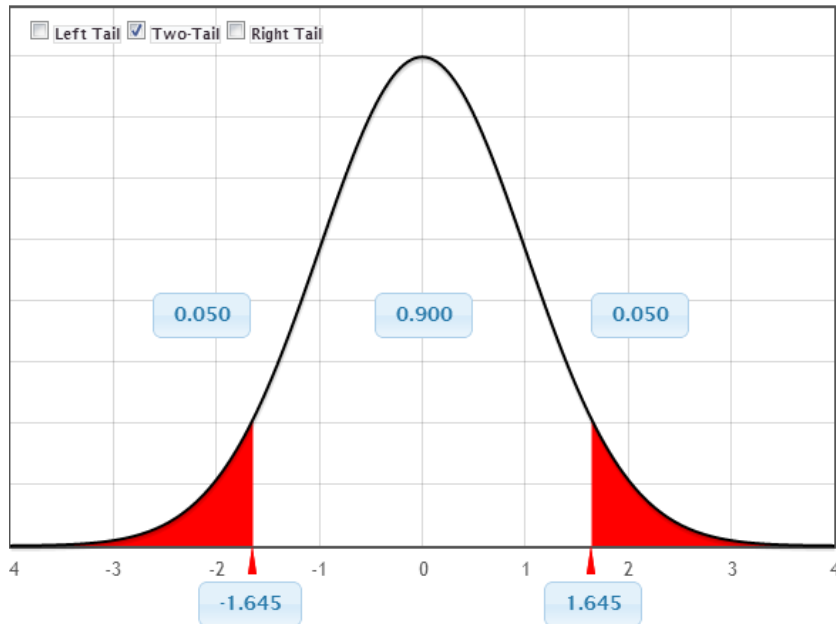
95% of all values fall within **1.96** SE's of the mean

```
> qnorm(.975)  
[1] 1.959964
```

What if we wanted a 90% CI? What z-score should we use to get the margin of error?

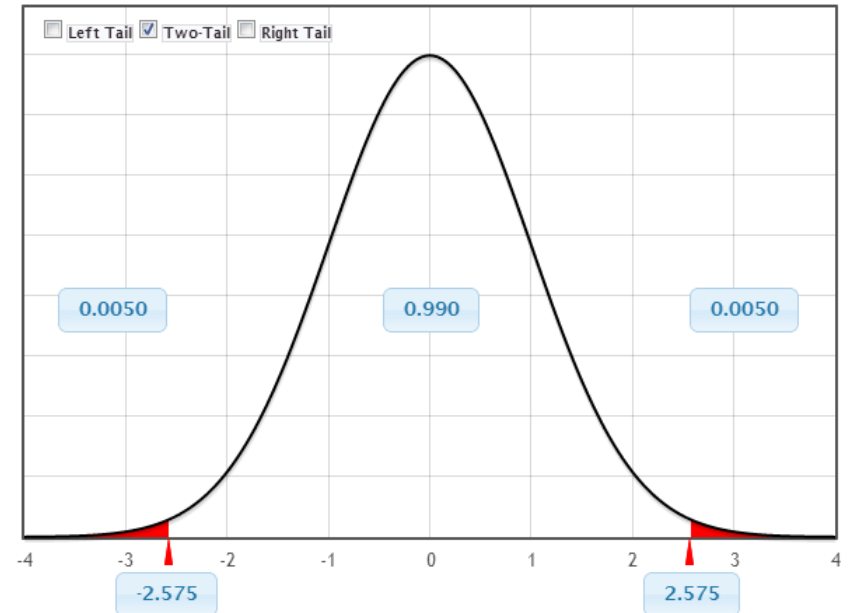
Other Levels of Confidence

www.lock5stat.com/statkey



90% Confidence
 $z^* = 1.645$

```
> qnorm(.95)  
[1] 1.644854
```



99% Confidence
 $z^* = 2.576$

```
> qnorm(.995)  
[1] 2.575829
```

Confidence Interval using $N(0,1)$

If a statistic is normally distributed, we find a confidence interval for the parameter using

$$\textit{statistic} \pm z^* SE$$

where the area between $-z^*$ and $+z^*$ in the standard normal distribution is the desired level of confidence.

Global Warming

What percentage of Americans believe in global warming?

A survey on 2,251 randomly selected individuals conducted in October 2010 found that 1328 answered “Yes” to the question

“Is there solid evidence of global warming?”

Give and interpret a 95% CI for the proportion of Americans who believe there is solid evidence of global warming.

Source: “Wide Partisan Divide Over Global Warming”, Pew Research Center, 10/27/10. s

Global Warming

Bootstrap For One Categorical Variable [\[Return to StatKey Index\]](#)

Custom Data ▾

Edit Data

Generate 1 Samples

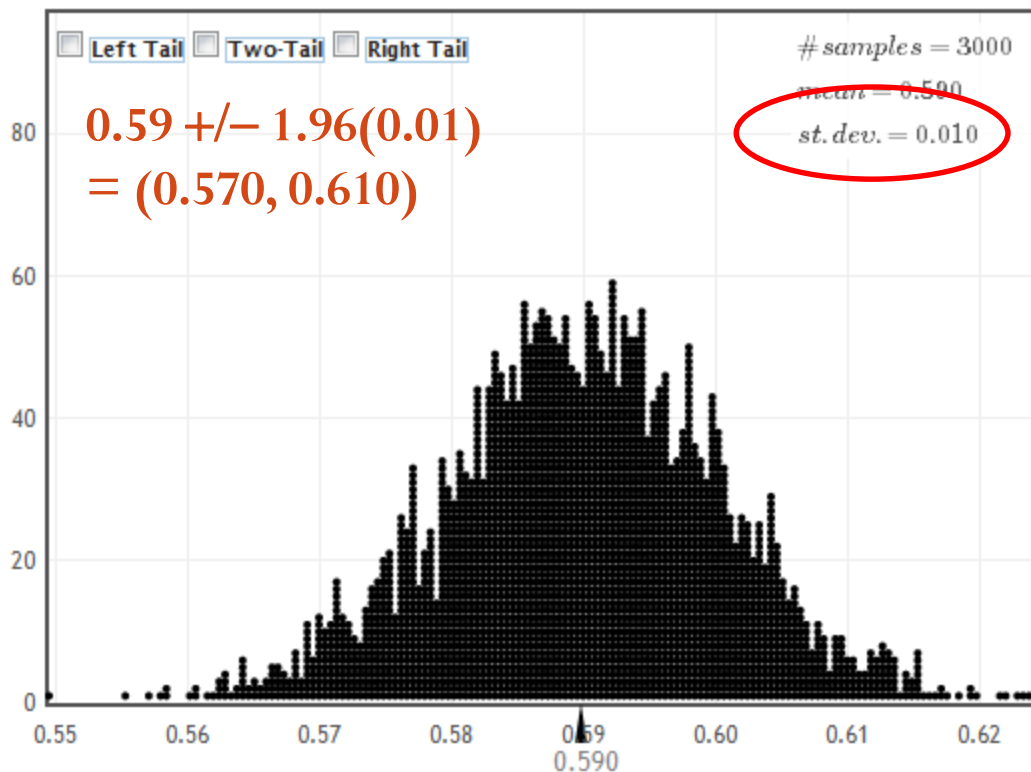
Generate 10 Samples

Generate 100 Samples

Generate 1000 Samples

Reset Plot

Bootstrap Dotplot of Proportion ▾



Original Sample

Count	n	Proportion
1328	2251	0.590

Bootstrap Sample

Count	n	Proportion
1304	2251	0.579

We are 95% confident that the true percentage of all Americans that believe there is solid evidence of global warming is between 57.0% and 61.0%

Global Warming

- What is a 90% confidence interval for the proportion of US adults who believe in global warming?

$z^* = 1.645$ for 90% confidence

$$0.59 \pm 1.645(0.01) = (0.574, 0.606)$$

- What is a 99% confidence interval?

$z^* = 2.576$ for 99% confidence

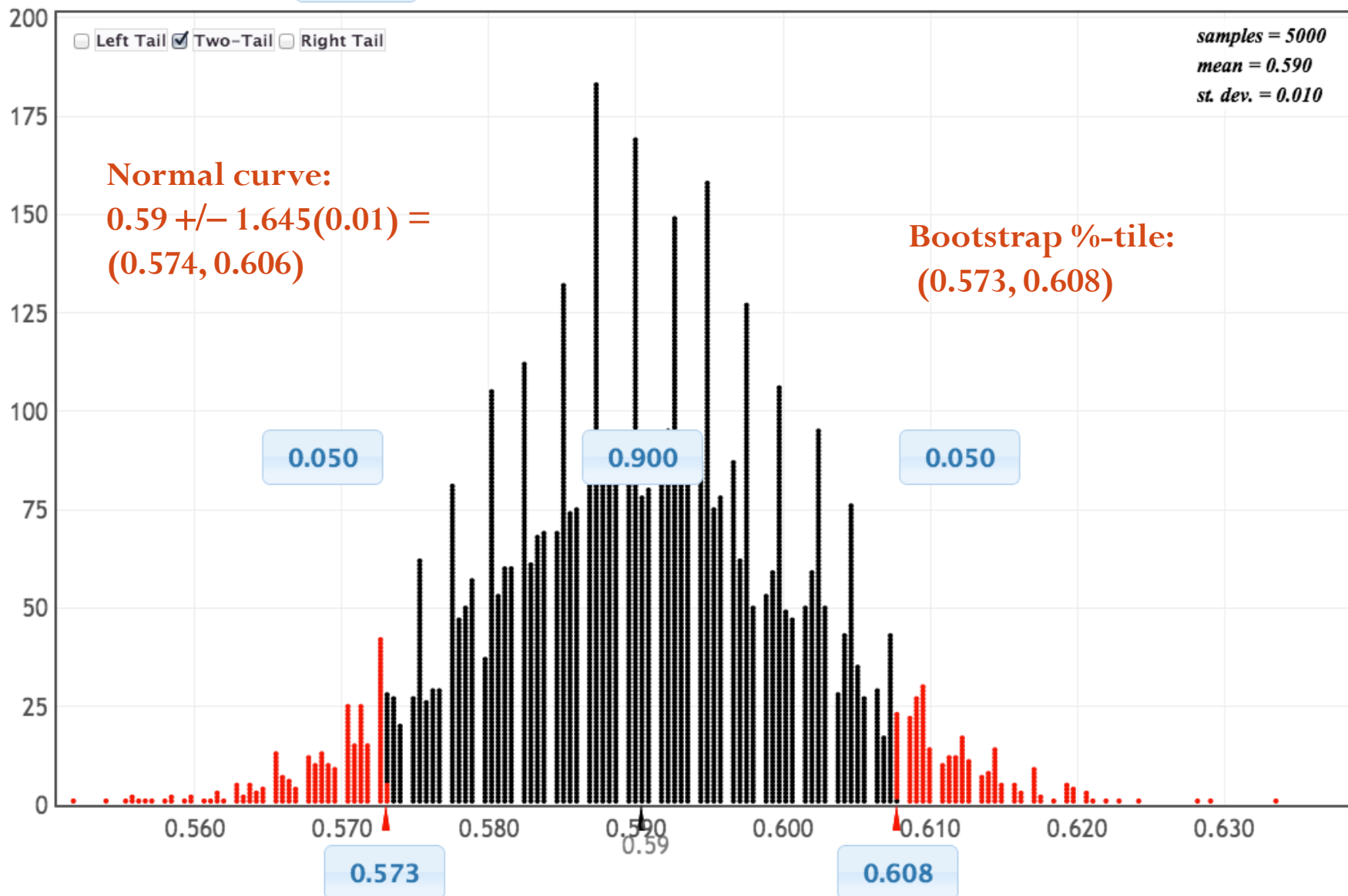
$$0.59 \pm 2.576(0.01) = (0.564, 0.616)$$

- Remember, more confidence = wider interval
- So how do these compare to the bootstrap CI?

Global Warming: 90% CI

Bootstrap Dotplot of

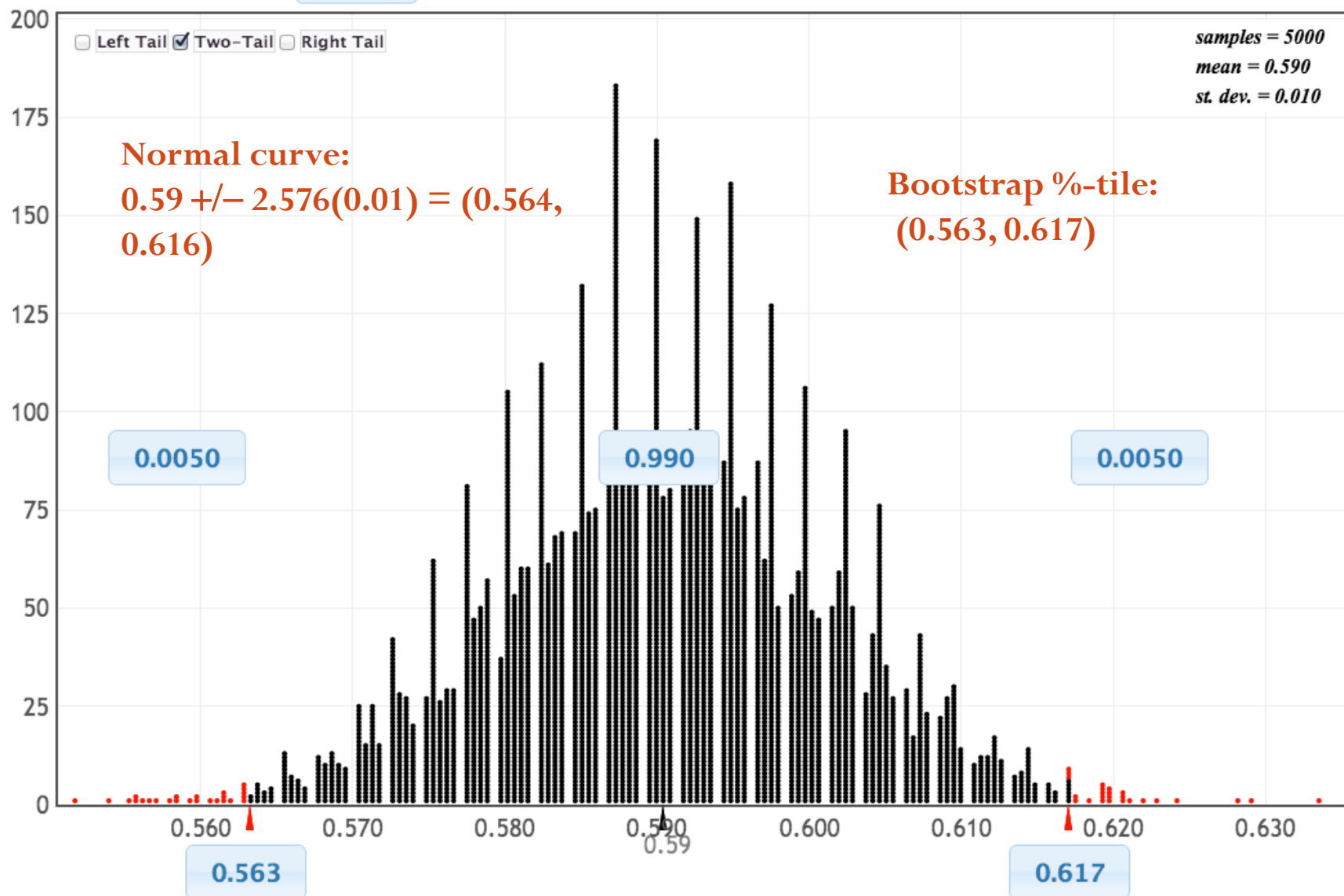
Proportion ▾



Global Warming: 99% CI

Bootstrap Dotplot of

Proportion ▾



Standard Error

- Wouldn't it be nice if we could compute the standard error *without* doing thousands of simulations?
- We can!!!
- Or rather, we'll be able to next class!