# Introduction to hypothesis tests

#### Overview

Questions about homework 4?

Last review of confidence intervals

A brief note on probability

Introduction to hypothesis tests

### Questions about homework 4?

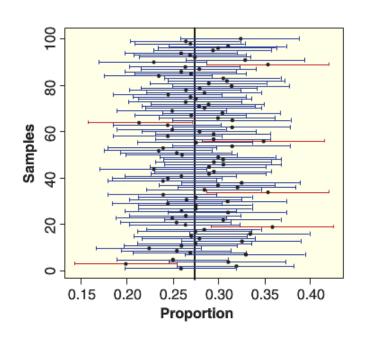
Homework 5 has been posted

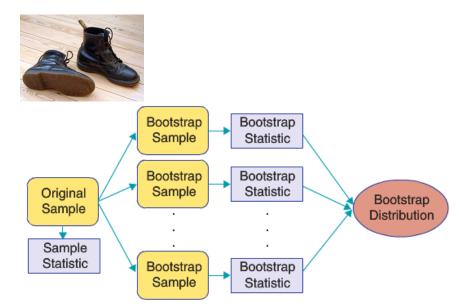
• Due on Gradescope at 11:30pm on Sunday February 23<sup>rd</sup>

### Are we feeling confident about confidence intervals?

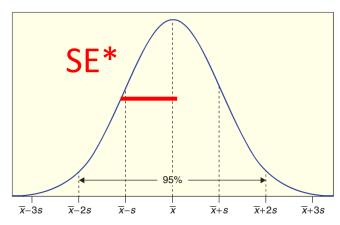
Confidence intervals give range of plausible values for a parameter





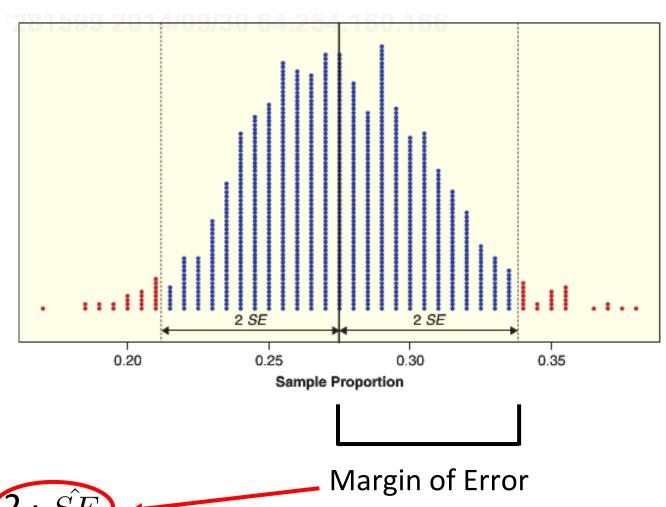


#### **Bootstrap distribution**



95% confidence interval: stat  $\pm 2 \cdot \hat{SE}$ 

### Are we feeling confident about confidence intervals?



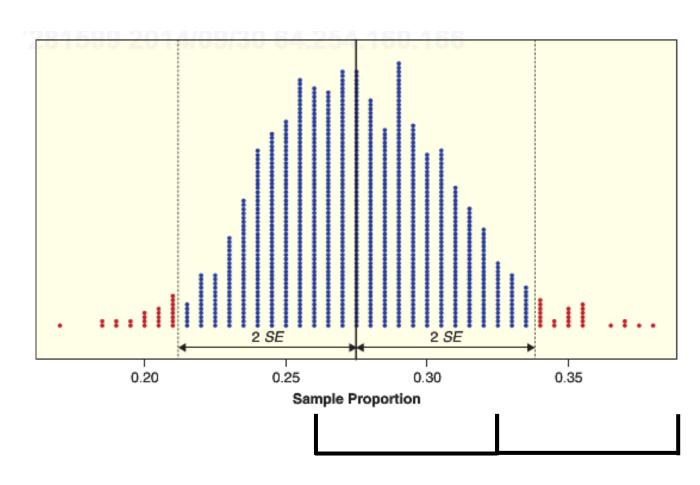
95% confidence interval: stat  $\pm (2 \cdot \hat{SE})$ 

Half width of confidence interval

### Are we feeling confident about confidence intervals?

How far can we slide the center of the confidence interval until we don't capture the parameter?

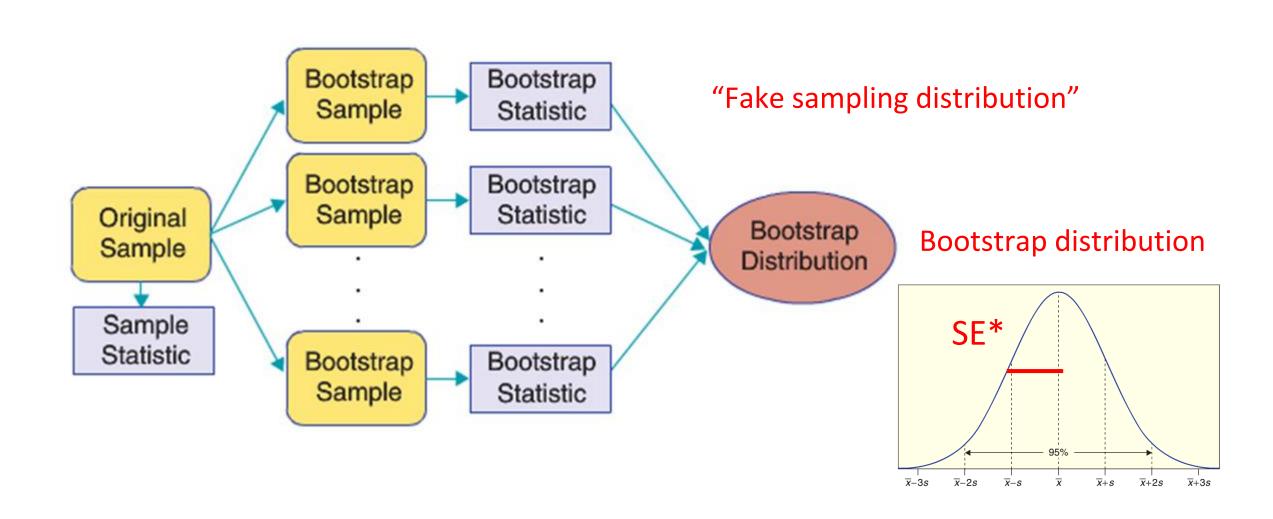
What would happen if we made the margin of error smaller?



Confidence interval

95% confidence interval: stat  $\pm 2 \cdot \hat{SE}$ 

# Using the bootstrap to estimate the standard error (SE\*)



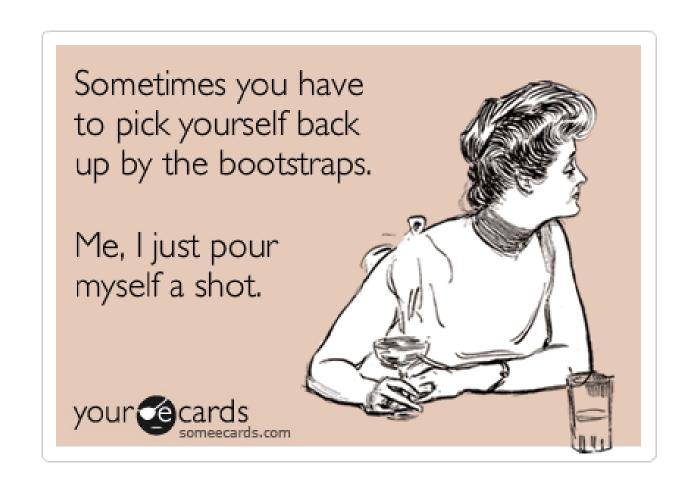
## Bootstrap distribution in R for μ

```
my_sample <- c(21, 29, 25, 19, 24, 22, 25, 26, 25, 29) # n = 10 points here
my_stat <- mean(my_sample) # x-bar</pre>
bootstrap_dist <- do_it(10000) * {
      curr_boot <- sample(my_sample , 10, replace = TRUE)</pre>
      mean(curr boot)
SE_boot <- sd(bootstrap_dist)</pre>
Cl <- c(my stat - 2 * SE boot,
                                     my stat + 2 * SE boot)
```

## Bootstrap distribution in R for $\pi$

```
my sample <- c("heads", "tails", "tails", "heads", "heads") # n = 5
my_stat <- sum(my_sample == "heads")/5 # p-hat
bootstrap_dist <- do it(10000) * {
      curr_boot <- sample(my_sample , 5, replace = TRUE)</pre>
       sum(my sample == "heads")/5
SE_boot <- sd(bootstrap_dist)</pre>
Cl <- c(my stat - 2 * SE boot, my stat + 2 * SE boot)
```

#### Any questions about confidence intervals or the bootstrap?



## A quick note on probability

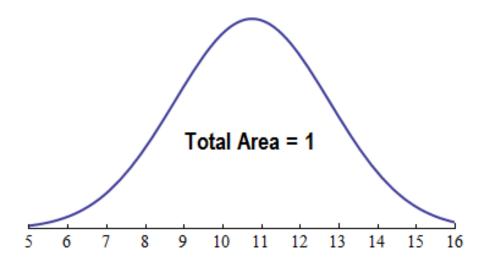
Probability is a way of measuring the likelihood that an event will occur

Probability models assigns a number between 0 and 1 to the outcome of an event occurring

We can use a probability model to calculate the probability of an event

#### For example:

- Pr(X < 11) = 0.55
- Pr(X > 20) = 0



# Statistical tests (hypothesis test)

A **statistical test** uses data from a sample to assess a claim about a population

Example 1: we might make the claim that Trump's approval rating for all all US citizens is 45%

How can we write this using symbols?

• 
$$\pi = .45$$

# Statistical tests (hypothesis test)

A **statistical test** uses data from a sample to assess a claim about a population

Example 2: we might make the claim that the average height of a baseball player is 72 inches

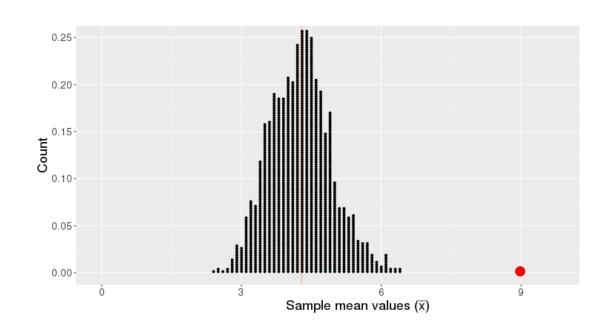
How can we write this using symbols?

• 
$$\mu = 72$$

# Basic hypothesis test logic

We start with a claim about a population parameter

This claim implies we should get a certain distribution of statistics



If our observed statistic is highly unlikely, we reject the claim

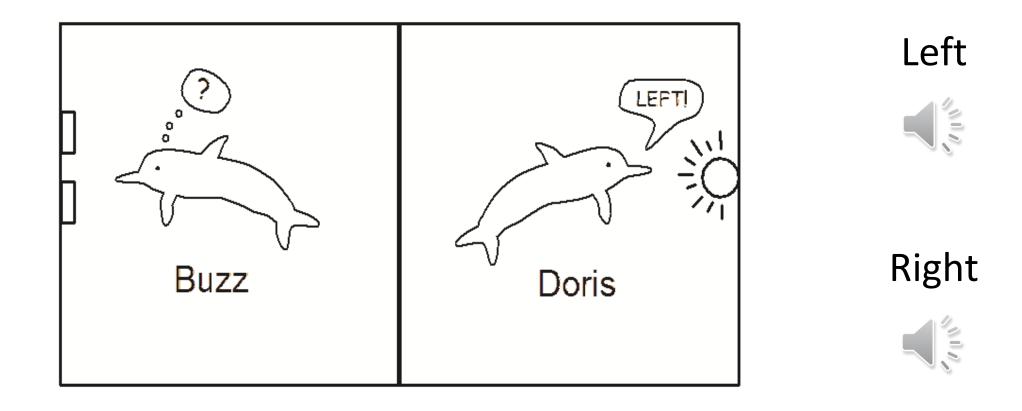
# Are dolphins capable of abstract communication?

Dr. Jarvis Bastian is the 1960's wanted to know whether dolphins are capable of abstract communication

He used an old headlight to communicate with two dolphins (Doris and Buzz)

- Stead light = push button on right to get food
- Flashing = push button on the left to get food

A canvas was then put in the middle of the pool with Doris on one side and buzz on the other



Buzz got 15 out of 16 trials correct

### Write down/discuss answers to the following questions

- 1. What are the cases here?
- 2. What is the variable of interest and is it categorical or quantitative?
- 3. What is the observed statistic and what symbols should we use to denote it?
- 4. What is the population parameter we are trying to estimate and what symbol should we use to denote it?
- 5. Do you think the results are due to chance?
  - i.e., how many correct answers do you think Buzz would have gotten if he was guessing?
- 6. Are dolphins capable of abstract communication?

# The dolphin communication study

If Buzz was just guessing, what would we expect the value of the parameter to be?

$$\pi = 0.5$$

If Buzz was not guessing, what would we expect the value of the parameter to be?

$$\pi > 0.5$$

#### Chance models

How can we assess whether 15 out of 16 correct trials ( $\hat{p}$  = .975) is beyond what we would expect to see by chance?

• i.e., beyond what we would expect to see if  $\pi = 0.5$ ?

If buzz was guessing we can model his guesses as a coin flip:

Heads = correct guess

Tails = incorrect guess

Let's flip a coin 16 times and see how many times we get 15 heads

#### Chance models

To really be sure, how many repetitions of flipping a coin 16 times should we do?

Any ideas how to do this?



#### Flipping coins using ClassTools functions

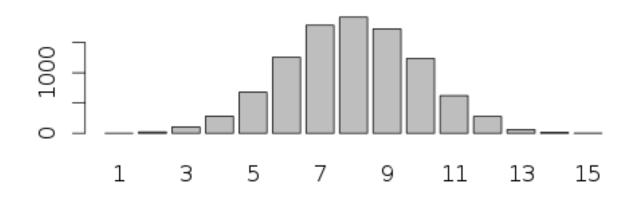
```
rflip_count() returns the number of heads out of num_flip coin flips:
  rflip_count(num_flips, prob = .5)
    num_flips: the number of times to flip a coin

    16 for Doris/Buzz

    prob: the probability of success on each trial
        • .5 if Buzz was guessing
We can repeat flips many times using the do it() function:
   library(ClassTools)
    flip_simulations <- do_it(10000) * {
        rflip count(16, prob = .5)
```

# Simulating Flipping 16 coins 10,000

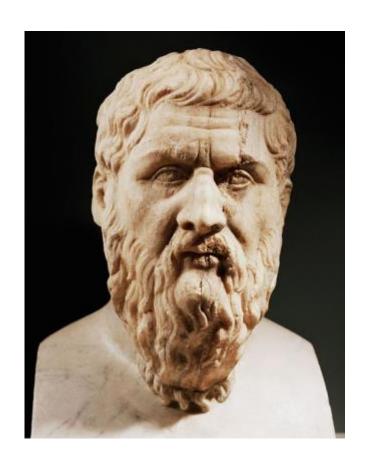
0	0
1	1
1 2	22
3	105
4	283
5 6	679
6	1257
7	1786
	1920
9	1726
10	1238
11	623
12	279
13	63
14	15
15	3
16	0



Is it likely that Buzz was guessing?

Are dolphins capable of abstract communication?

### Question: who is this?



A: Gorgias

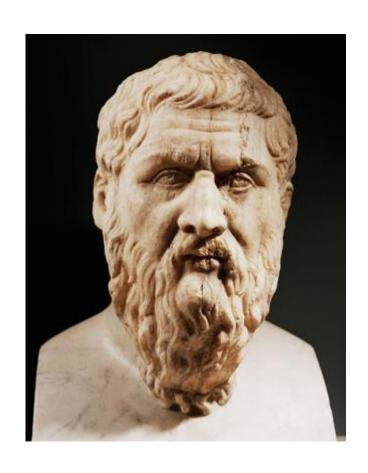
Question: Who is Gorgias?

A: a skeptic

Question: Does Gorgias believe Doris and Buzz can communicate?

A: No!

# Question: who is this?



Gorgias believes in the *null hypothesis* 

- that Buzz was guessing

How can we write the null hypothesis in symbols?

$$H_0$$
:  $\pi = 0.5$ 

We believe in the *alternative hypothesis* 

- Doris an Buzz can communicate

How can we write the alternative hypothesis in symbols?

$$H_{\Delta}$$
:  $\pi > 0.5$ 

### Question: who is this?

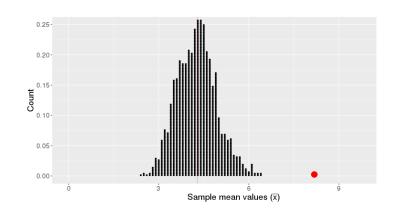
To prove Gorgias wrong, we will start by assuming he is right!

Namely, we will assume  $H_0$  that  $\pi = 0.5$ )

We will then generate a number of statistics ( $\hat{p}$ ) that are consistent with H<sub>0</sub>

• i.e., we will create a *null distribution* 

If our observed statistic looks very different from the statistics generated under we can reject H<sub>0</sub> and accept H<sub>A</sub>

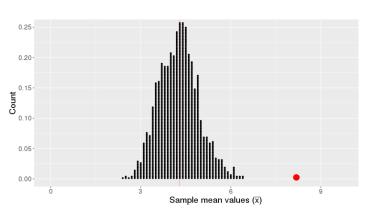


## Terminology

Null Hypothesis (Claim that there is no effect or no difference

**Alternative Hypothesis** (H<sub>a</sub>): Claim for which we seek significant evidence

The alternative hypothesis is established by observing evidence that inconsistent with the null hypothesis



# Review: the dolphin communication study

- 1. What is the null hypothesis?
- 2. We can write this in terms of the population parameter as:

$$H_0$$
:  $\pi = 0.5$ 

3. What is the alternative hypothesis?

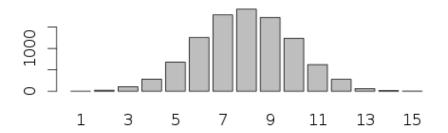
$$H_{\Delta}$$
:  $\pi > 0.5$ 

#### Null Distribution

A **null distribution** is the distribution of statistics one would expect if the null hypothesis  $(H_0)$  was true

i.e., the null distribution is the statistics one would expect to get if nothing interesting was happening

note: the book calls this the randomization distribution



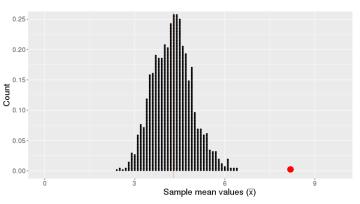
#### P-values

A **p-value** is the probability, of obtaining a statistic as as (or more) extreme than the observed sample *if the null hypothesis was true* 

• i.e., the probability that we would get a statistic as extreme as our <u>observed</u> statistic from the <u>null distribution</u>

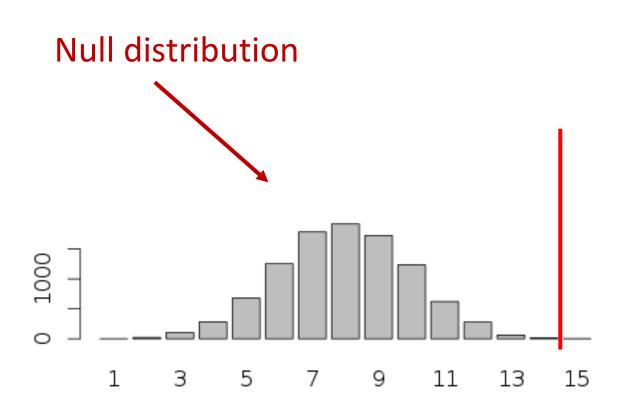
 $Pr(STAT \ge observed statistic \mid H_0 = True)$ 

The smaller the p-value, the stronger the statistic evidence is against the null hypothesis



## Buzz and Doris example

0	0
1	1
2	22
1 2 3 4	105
4	283
5	679
6 7	1257
7	1786
8	1920
9	1726
10	1238
11	623
12	279
13	63
14	15
15	3
16	0



$$p$$
-value =  $3/10000 = 0.0003$ 

# Statistical significance

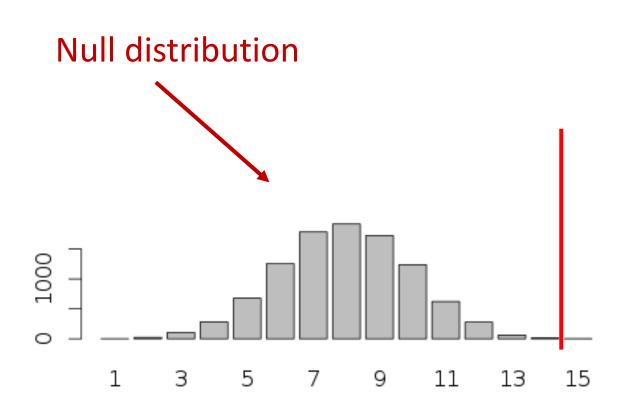
When our observed sample statistic is unlikely to come from the null distribution, we say the sample results are **statistically significant** 

• i.e., we have a small p-value

'Statistically significant' results mean we have strong evidence against  $H_0$  in favor of  $H_a$ 

## Buzz and Doris example

0	0
1	1
1 2 3 4	22
3	105
4	283
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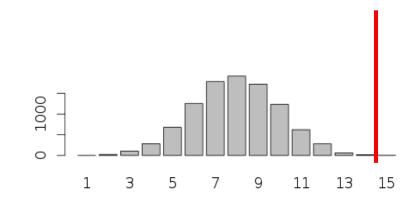


$$p$$
-value =  $3/10000 = 0.0003$ 

### Getting p-values using ClassTools functions

#### Flipping coints many times:

```
flip_simulations <- do_it(10000) * {
    rflip_count(16, prob = .5)
}</pre>
```



We can get the number of values as or more extreme than an observed statistic (obs\_stat) using the pnull() function:

```
obs_stat <- ?
p_value <- pnull(obs_stat, flip_simulations, lower.tail = FALSE)</pre>
```

#### Key steps hypothesis testing

#### 1. State the null hypothesis... and the alternative hypothesis

- Buzz is just guessing so the results are due to chance:  $H_0$ :  $\pi = 0.5$
- Buzz is getting more correct results than expected by chance:  $H_A$ :  $\pi > 0.5$

#### 2. Calculate the observed statistic

• Buzz got 15 out of 16 guesses correct, or  $\hat{p}$  = .973

#### 3. Create a null distribution that is consistent with the null hypothesis

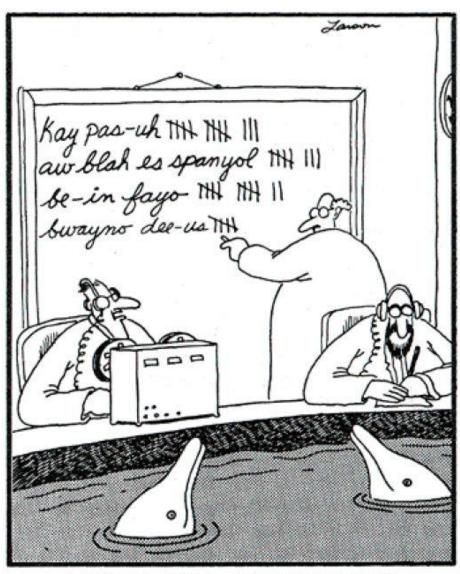
• i.e., what statistics would we expect if Buzz was just guessing

#### 4. Examine how likely the observed statistic is to come from the null distribution

- What is the probability that the dolphins would guess 15 or more correct?
- i.e., what is the p-value

#### 5. Make a judgement

- If we have a small p-value, this means that  $\pi = .5$  is unlikely and so  $\pi > .5$
- i.e., we say our results are 'statistically significant'



"Matthews ... we're getting another one of those strange 'aw blah es span yol' sounds."