

# Lecture 15: Hypothesis Testing Part II

## Chapter 4.3

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There are two potential outcomes of a hypothesis test. Either we

- ▶ reject  $H_0$
- ▶ fail to reject  $H_0$

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	$H_A$ true	Type II Error	OK

Two kinds of errors:

- ▶ Type I Error: a false positive (test result)
- ▶ Type II Error: a false negative (test result)

## Type I Errors: US Criminal Justice System

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In this case:

- ▶ Type I error is putting an innocent person in jail (considered worse)
- ▶ Type II error is letting a guilty person go free.

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Failing to reject  $H_0$  when  $H_A$  is true is not “patting down” passenger X when they have a weapon.

Hence the long lines at airport security.

# Goals for Today

- ▶ Define significance level
- ▶ Tie-in p-Values with sampling distributions
- ▶ Example



# Significance Level

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Say you flip a coin you think is fair 1000 times. Say you observe

- ▶ 501 heads? Do you think the coin is biased?
- ▶ 525 heads? Do you think the coin is biased?
- ▶ 900 heads? Do you think the coin is biased?

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Note the p-value is different than the population proportion  $p$  (bad historical choice).

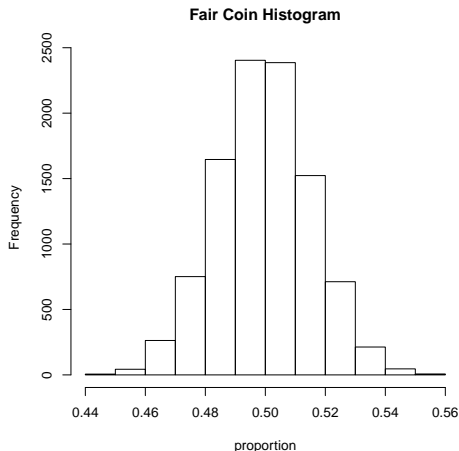


# p-Values

# Thought experiment: Coin Flips

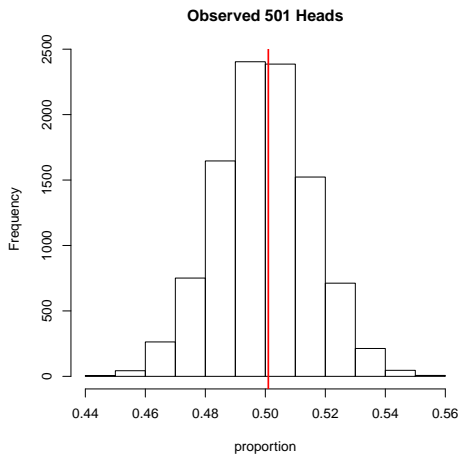
# Sampling Distribution of $\hat{p}$

Under  $H_0$  that the coin is fair, i.e.  $p = p_0 = 0.5$ , the sampling distribution of  $\hat{p}$  when  $n = 1000$  is:



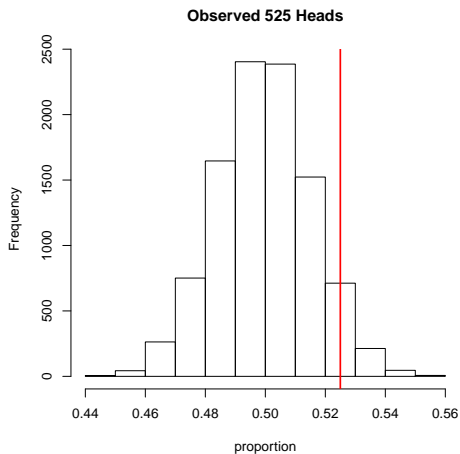
Say we observe...

$$\hat{p} = \frac{501}{1000}$$



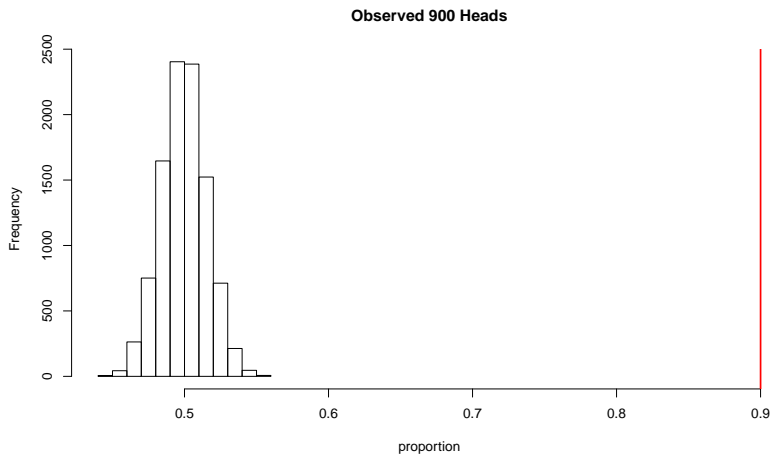
Say we observe...

$$\hat{p} = \frac{525}{1000}$$



Say we observe...

$$\hat{p} = \frac{900}{1000}$$



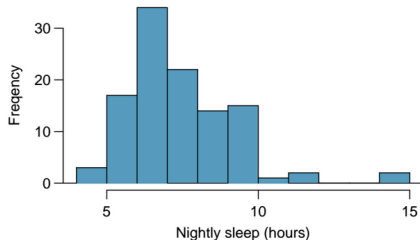
## Example about Sleep Habits

A poll found that college students sleep about 7 hours a night. Researchers suspect that Reedies sleep more. They want to investigate this claim at a pre-specified  $\alpha = 0.05$  level.

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They sample  $n = 110$  Reedies and find that  $\bar{x} = 7.42$  and  $s = 1.75$  and the histogram looks like:





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Since the p-value  $0.007 < 0.05 = \alpha$ , the pre-specified significance level, it has a high degree of extremeness, and thus we reject  $H_0$ .

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**Conclusion:** we reject (at the  $\alpha = 0.05$  significance level) the hypothesis that the average # of hours of Reedies sleep is 7, in favor of the hypothesis that sleep more.

## Example about Sleep Habits

**Correct interpretation of the p-value:** If the null hypothesis is true ( $\mu = 7$ ), the probability of observing a sample mean  $\bar{x} = 7.42$  or greater is 0.007.

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**Incorrect interpretation of the p-value:** The probability that the null hypothesis ( $\mu = 7$ ) is true is 0.007.

## Next Time

- ▶ How big a sample size do I need? i.e. power calculations
- ▶ Statistical vs practical significance