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JOHNS HOPKINS  
BLOOMBERG  
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## Section D

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The p-Value *in Even More Detail!*

# p-Values

- p-values are probabilities (numbers between 0 and 1)
- Small p-values mean that the sample results are unlikely when the null is true
- The p-value is the probability of obtaining a result as extreme or more extreme than you did by chance alone assuming the null hypothesis  $H_0$  is true
  - How likely your sample result (and other result less likely) are if null is true

# p-Values

- The p-value is not the probability that the null hypothesis is true!
- The p-value alone imparts no information about scientific/ substantive content in result of a study
- Example: from Example 3, the researchers found a statistically significant ( $p=0.005!$ ) difference in average LDL cholesterol levels in men who had been on a diet including corn flakes versus the same men on a diet including oat bran cereal
  - Which diet showed lower average LDL levels?
  - How much was the difference; does it mean anything nutritionally?

# p-Values

- If the p-value is small either a very rare event occurred and
  - $H_0$  is true
  - or
  - $H_0$  is false
- Type I error
  - Claim  $H_A$  is true when in fact  $H_0$  is true
  - The probability of making a Type I error is called the *alpha-level ( $\alpha$ -level)* or *significance level*

## Note on the p-Value and the Alpha-Level

- If the p-value is less than some pre-determined cutoff (e.g., .05), the result is called *statistically significant*
- This cutoff is the  *$\alpha$ -level*
  - $\alpha$ -level is the probability of a type I error
  - It is the probability of falsely rejecting  $H_0$  when  $H_0$  true
- Idea: to keep the chance of “making a mistake” when the  $H_0$  is true low and only reject if the sample result is “unlikely”
  - Unlikeliness threshold is determined by  *$\alpha$ -level*

# Note on the p-Value and the Alpha-Level

- Truth versus decision made by hypothesis testing

		TRUTH	
		$H_0$	$H_A$
Reject $H_0$	Type I Error alpha-level	Power 1-beta	
Not Reject $H_0$		Type II Error beta	

# Note on the p-Value and the Alpha-Level

- Truth versus decision made by hypothesis testing

		TRUTH	
		$H_0$	$H_A$
Decision	Reject $H_0$	Type I Error alpha-level	Power 1-beta
	Not Reject $H_0$		Type II Error beta



# Note on the p-Value and the Alpha-Level

- Truth versus decision made by hypothesis testing

		TRUTH	
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Reject $H_0$	Type I Error alpha-level	Power 1-beta	
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# Note on the p-Value and the Alpha-Level

- Truth versus decision made by hypothesis testing

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## More on p-Value: One-Sided vs. Two-Sided Controversy

- Two-sided p-value (BP/OC:  $p = .009$ )
  - Probability of a result as or more extreme than observed (either positive or negative)
- One-sided p-value
  - Probability of a more extreme positive result than observed or a more extreme negative result: only considers extremes in one direction of null when evaluation how likely your sample result is (and results less likely)
  - If the direction of the alternative hypothesis in the one-sided test is the same as the direction of the sample result in terms of above/below the null, then the one-sided p-value will be half the two-sided p-value

# Stata Output

- One-sided alternative: true mean difference  $>0$ 
  - Sample mean difference was greater than 0

```
. ttesti 10 4.8 4.6 0
```

```
One-sample t test
```

```
-----  
      |      Obs      Mean      Std. Err.      Std. Dev.      [95% Conf. Interval]  
-----+-----  
      x |      10      4.8      1.454648      4.6      1.509358      8.090642  
-----
```

```
      mean = mean(x)                                t =      3.2998  
Ho: mean = 0                                       degrees of freedom =      9
```

```
      Ha: mean < 0  
Pr(T < t) = 0.9954
```

```
      Ha: mean != 0  
Pr(|T| > |t|) = 0.0092
```

```
      Ha: mean > 0  
Pr(T > t) = 0.0046
```

# Stata Output

- One-sided alternative: true mean difference  $< 0$ 
  - Sample mean difference was greater than 0

```
. ttesti 10 4.8 4.6 0
```

```
One-sample t test
```

```
-----  
      |      Obs      Mean   Std. Err.   Std. Dev.   [95% Conf. Interval]  
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## More on the p-Value

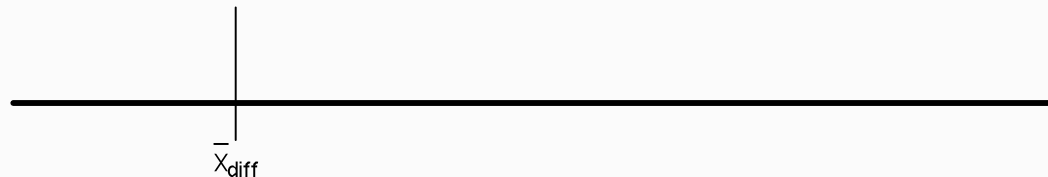
- In some cases, a one-sided alternative may not make scientific sense
  - In the absence of pre-existing information, in evaluating the BP/OC relationship, wouldn't either result be interesting and useful? (i.e., negative or positive association?)
- In some cases, a one-sided alternative often makes scientific sense
  - For example: not really interested if new treatment is worse than old treatment—only care whether it's better
- However: because of “culture of p-value” and sanctity of “.05,” one-sided p-values are viewed with suspicion
- In this course, we will use two-sided p-values exclusively

## Connection: Hypothesis Testing and CIs

- The confidence interval gives plausible values for the population parameter
  - “Data take me to the truth”
- Hypothesis testing postulates two choice for the population parameter
  - “Here are two possibilities for the truth; data help me choose one”

# 95% Confidence Interval

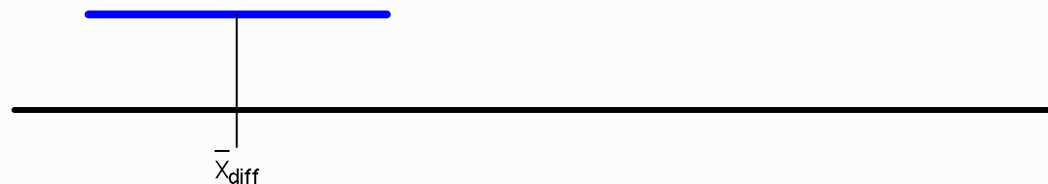
- If 0 is not in the 95% CI, then we would reject  $H_0$  that  $\mu = 0$  at level  $\alpha = .05$  (the p-value  $< .05$ )
- Why?
- With confidence interval we start at sample mean difference and go two standard errors in either direction (or slightly more in small samples)





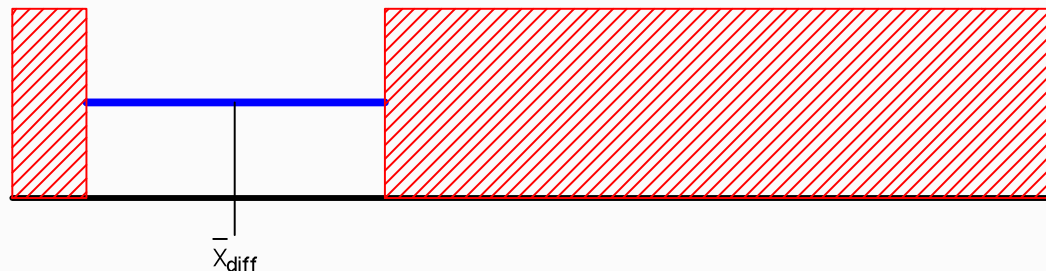
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- Why?
- With confidence interval we start at sample mean difference and go two standard errors in either direction (or slightly more in small samples)



# 95% Confidence Interval

- If 0 is not in the 95% CI, then this must mean  $\bar{x}$  is  $> 2$  standard errors away from 0 (either above or below)
- Hence, the distance ( $t$ ) will be  $> 2$  or  $< -2$ : and the resulting p-value  $< .05$



## 95% Confidence Interval and p-Value

- In the BP/OC example, the 95% confidence interval tells us that the p-value is less than .05, but it doesn't tell us that it is  $p = .009$
- The confidence interval and the p-value are complementary
- However, you can't get the exact p-value from just looking at a confidence interval, and you can't get a sense of the scientific/substantive significance of your study results by looking at a p-value

## More on the p-Value

- Statistical significance does not imply/prove causation
- For example: in the blood pressure/oral contraceptives example, there could be other factors that could explain the change in blood pressure
- A significant p-value is only ruling out random sampling (chance) as the explanation
- Need a comparison group to better establish causality
  - Self-selected (may be okay)
  - Randomized (better)

## More on the p-Value

- *Statistical significance* is not the same as *scientific significance*
- Hypothetical example: blood pressure and oral contraceptives:
  - Suppose:
    - ▶  $n = 100,000$ ;  $\bar{x}_{diff} = .03$  mmHg;  $s = 4.6$  mmHg
    - ▶ p-value = .04
- Big  $n$  can sometimes produce a small p-value, even though the magnitude of the effect is very small (not scientifically/substantively significant)
- Very important
  - Always report a confidence interval
  - 95% CI: 0.002-0.058 mmHg

## More on the p-Value

- *Lack of statistical significance* is not the same as *lack of scientific significance*
  - Must evaluate in context of study, sample size
- Small  $n$  can sometimes produce a non-significant even though the magnitude of the association at the population level is real and important (our study just can't detect it)
- *Low power* in small sample studies makes not rejecting hard to interpret
- Sometimes small studies are designed without power in mind just to generate preliminary data