

# Statistical significance using $p$ -value

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# Learning Objectives

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- (1) Able to understand the concepts of statistical inference and statistical significance.
  - (2) Able to apply the concept of statistical significance (p-value) in analyzing the data.
  - (3) Able to interpret the concept of statistical significance (p-value) in making valid conclusions.

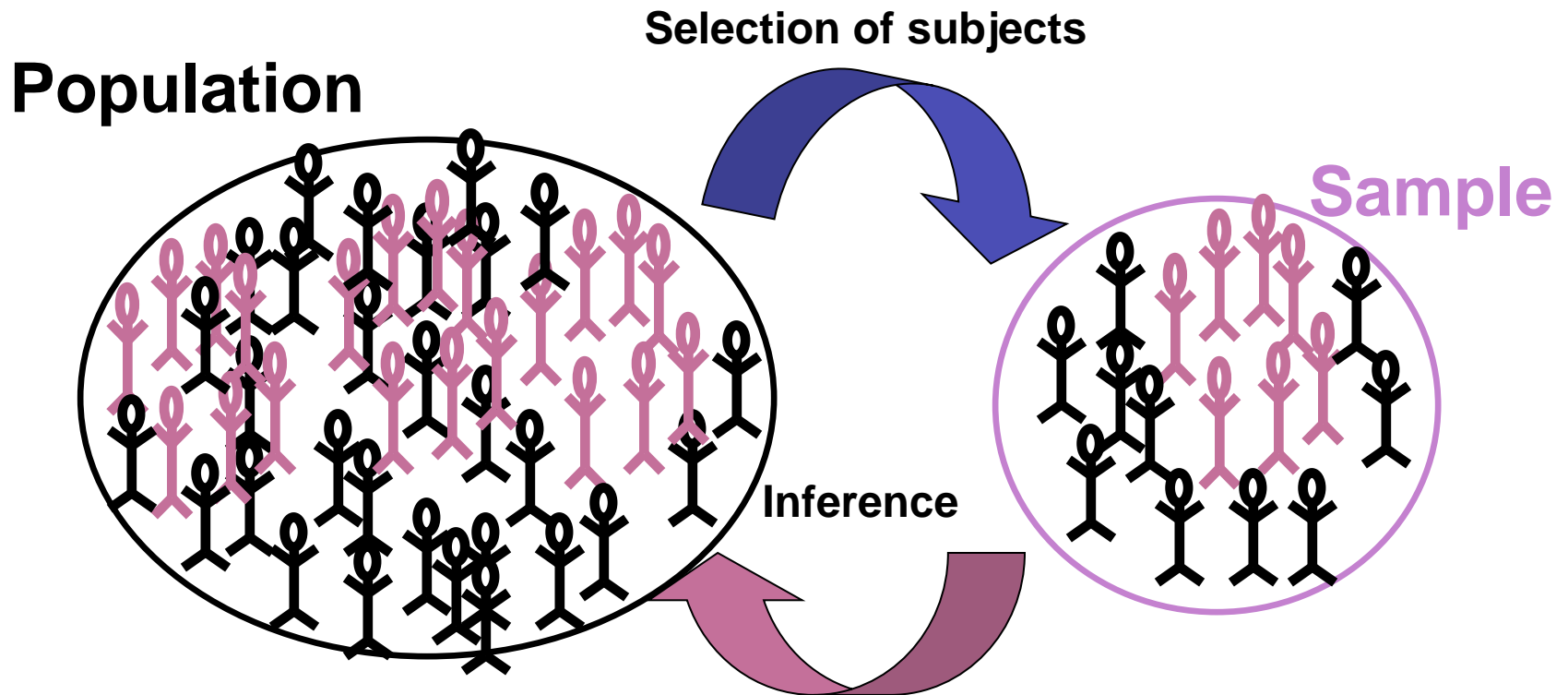
# Why use inferential statistics at all?

Average height of all 25-year-old men (**population**) in KSA is a **PARAMETER.**

The height of the members of a **sample** of 100 such men are measured; the average of those 100 numbers is a **STATISTIC.**

Using inferential statistics, we make inferences about population (taken to be unobservable) based on a random sample taken from the population of interest.

# Is risk factor X associated with disease Y?

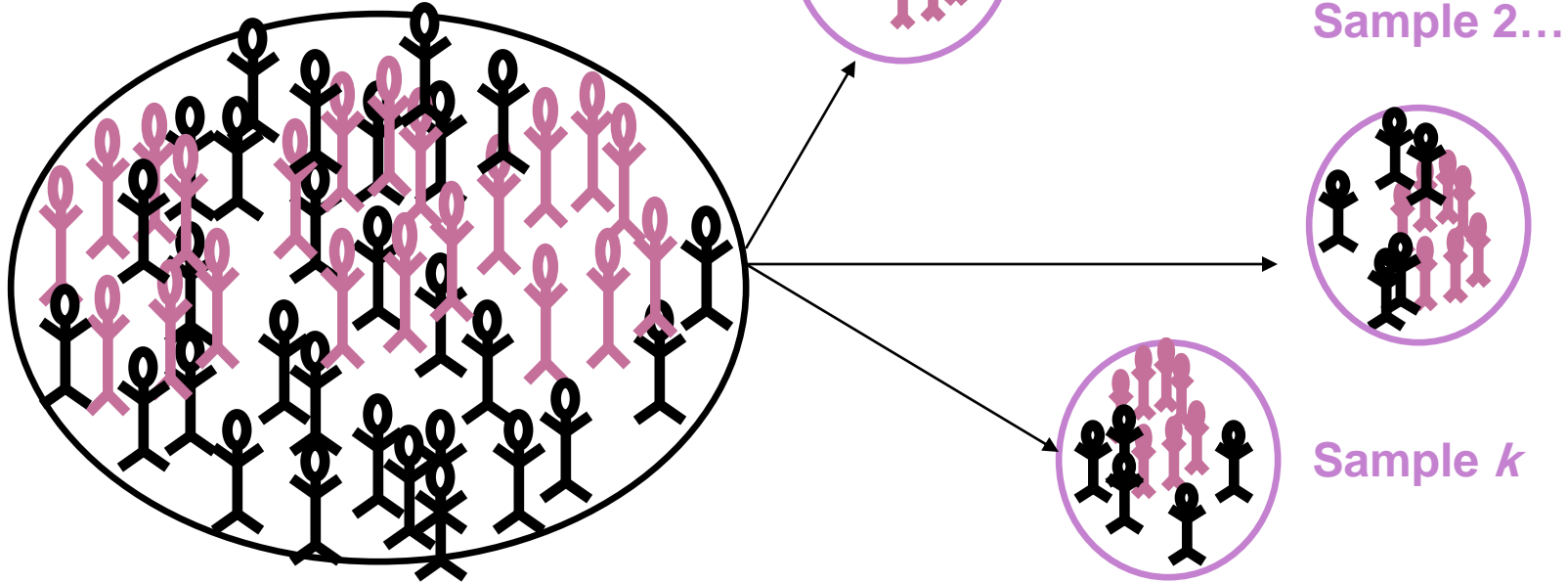


From the sample, we compute an estimate of the effect of X on Y (e.g., risk ratio if cohort study):

- Is the effect real? Did chance play a role?

# Why worry about chance?

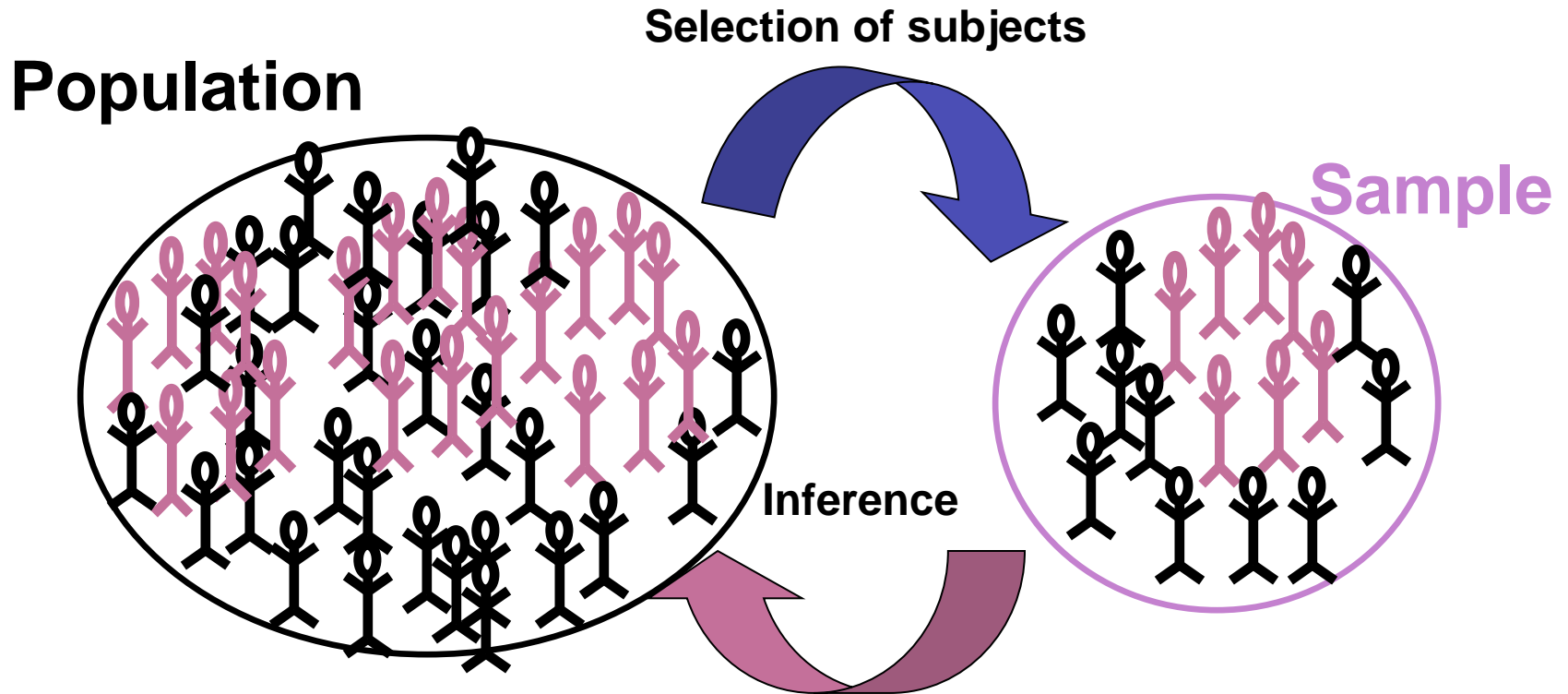
**Population**



**Sampling variability...**

**- you only get to pick one sample!**

# Interpreting the results



**Make inferences from data collected using laws of probability and statistics**

- tests of significance (p-value)
- confidence intervals

# Significance testing

- The interest is generally in comparing two groups (e.g., risk of outcome in the treatment and placebo group)
- The statistical test depends on the type of data and the study design



# Hypothesis Testing

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- *Null Hypothesis*

- There is no association between the predictors(associated factors) and outcome variable in the population
- Assuming there is no association, statistical tests estimate the probability that the association is due to chance

- *Alternate Hypothesis*

- The proposition that there is an association between the predictors and outcome variable
- We do not test this directly but accept it by default if the statistical test rejects the null hypothesis



## The Null and Alternative Hypothesis

- States the assumption (numerical) to be tested
  - Begin with the assumption that the null hypothesis is TRUE
  - Always contains the '=' sign
- The null hypothesis,  $H_0$
- 

The alternative hypothesis,  $H_a$

:

- Is the opposite of the null hypothesis
- Challenges the status quo
- Never contains just the '=' sign
- Is generally the hypothesis that is believed to be true by the researcher

## One and Two Sided Tests

- Hypothesis tests can be one or two sided (tailed)

- One tailed tests are directional:

$$H_0: \mu_1 - \mu_2 = 0$$

$$H_A: \mu_1 - \mu_2 > 0 \text{ or } H_A: \mu_1 - \mu_2 < 0$$

- Two tailed tests are not directional:

$$H_0: \mu_1 - \mu_2 = 0$$

$$H_A: \mu_1 - \mu_2 \neq 0$$

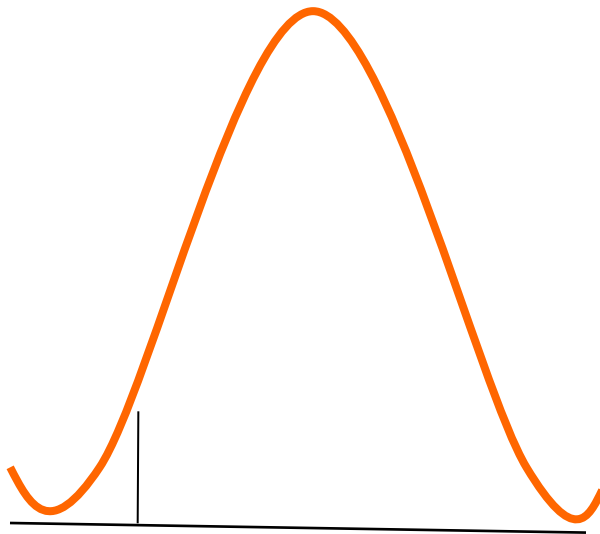
## When To Reject $H_0$ ?

Rejection region: set of all test statistic values for which  $H_0$  will be rejected

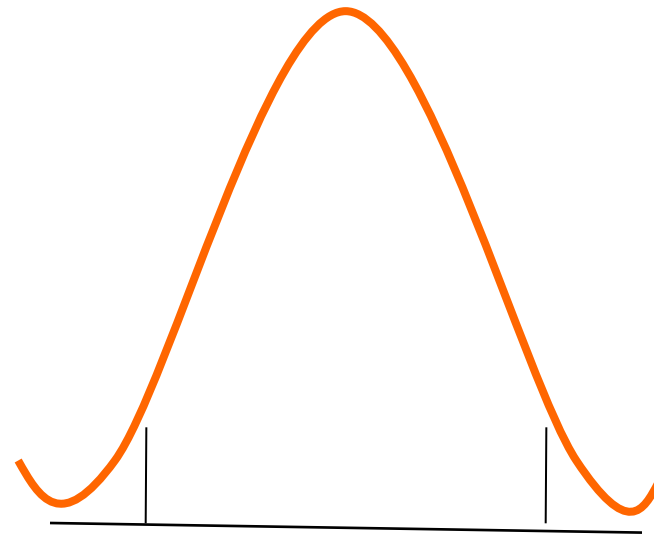
Level of significance,  $\alpha$ : Specified before an experiment to define rejection region

One Sided :  $\alpha = 0.05$

Two Sided:  $\alpha/2 = 0.025$



Critical Value = -1.64



Critical Values = -1.96 and +1.96



# Type-I and Type-II Errors

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- ❖  $\alpha$  = Probability of rejecting  $H_0$  when  $H_0$  is true
- ❖  $\alpha$  is called **significance level** of the test
- ❖  $\beta$  = Probability of not rejecting  $H_0$  when  $H_0$  is false
- ❖  $1-\beta$  is called **statistical power** of the test

# Diagnosis and statistical reasoning

		Disease status	
		Present	Absent
Test result	+ve	True +ve (sensitivity)	False +ve
	-ve	False -ve	True -ve (Specificity)

		<u>Significance Difference is</u>	
		Present	Absent
		(Ho <i>not</i> true)	(Ho is true)
<u>Test result</u>	Reject Ho	No error $1-\beta$	Type I err. $\alpha$
	Accept Ho	Type II err. $\beta$	No error $1-\alpha$

$\alpha$  : significance level

$1-\beta$  : power

# Significance testing

Subjects with Acute MI

Mortality  
IV nitrate

$P_N$

?  
<

Mortality  
No nitrate

$P_C$

- Suppose we do a clinical trial to answer the above question
- Even if IV nitrate has no effect on mortality, due to sampling variation, it is very unlikely that  $P_N = P_C$
- Any observed difference b/w groups may be due to treatment or a coincidence (or chance)

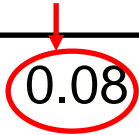
# Null Hypothesis( $H_0$ )

- There is no association between the independent and dependent/outcome variables
  - Formal basis for hypothesis testing
- In the example,  $H_0$  : "The administration of IV nitrate has no effect on mortality in MI patients" or  $P_N - P_C = 0$

# Obtaining $P$ values

Trial	Number dead / randomized		Risk Ratio	95% C.I.	P value
	Intravenous nitrate	Control			
Chiche	3/50	8/45	0.33	(0.09,1.13)	0.08
Bussman	4/31	12/29	0.24	(0.08,0.74)	0.01
Flaherty	11/56	11/48	0.83	(0.33,2.12)	0.70
Jaffe	4/57	2/57	2.04	(0.39,10.71)	0.40
Lis	5/64	10/76	0.56	(0.19,1.65)	0.29
Jugdutt	24/154	44/156	0.48	(0.28, 0.82)	0.007

How do we get this  $p$ -value?





# Example of significance testing

- In the Chiche trial:
  - $p_N = 3/50 = 0.06$ ;  $p_C = 8/45 = 0.178$
- Null hypothesis:
  - $H_0: p_N - p_C = 0$  or  $p_N = p_C$
- Statistical test:
  - Two-sample proportion

# Test statistic for Two Population Proportions

The test statistic for  $p_1 - p_2$  is a Z statistic:

$$Z = \frac{(p_N - p_C) - (P_N - P_C)_0}{\sqrt{\bar{p}(1 - \bar{p}) \left( \frac{1}{n_N} + \frac{1}{n_C} \right)}}$$

Observed difference

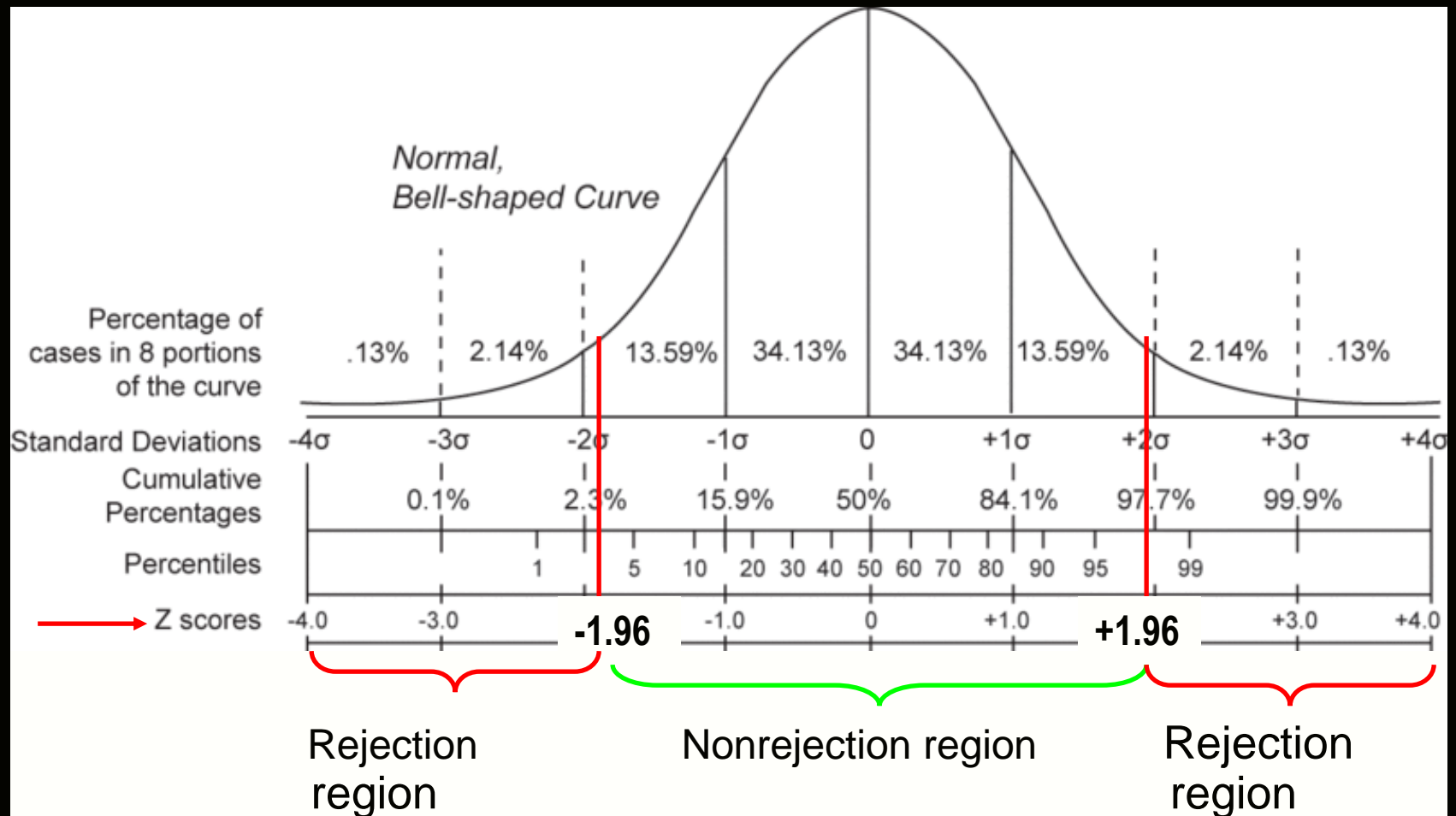
0  
Null hypothesis

No. of subjects in IV  
nitrate group

No. of subjects in  
control group

where  $\bar{p} = \frac{X_N + X_C}{n_N + n_C}$ ,  $p_N = \frac{X_N}{n_N}$ ,  $p_C = \frac{X_C}{n_C}$

# Testing significance at 0.05 level



$$Z_{\alpha/2} = 1.96$$

Reject  $H_0$  if  $Z < -Z_{\alpha/2}$  or  $Z > Z_{\alpha/2}$

# Two Population Proportions

*(continued)*

$$Z = \frac{(0.06 - 0.178)}{\sqrt{0.116(1 - 0.116) \left( \frac{1}{50} + \frac{1}{45} \right)}} = -1.79$$

where  $\bar{p} = \frac{3+8}{45+50} = 0.116$  ,  $p_N = \frac{3}{45} = 0.06$  ,  $p_C = \frac{8}{50} = 0.178$

# Statistical test for $p_1 - p_2$

Two Population Proportions, Independent Samples

$$Z = \frac{(0.06 - 0.178)}{\sqrt{0.116(1 - .116)\left(\frac{1}{50} + \frac{1}{45}\right)}} = -1.79$$

Since  $-1.79$  is  $>$  than  $-1.96$ , we fail to reject the null hypothesis.

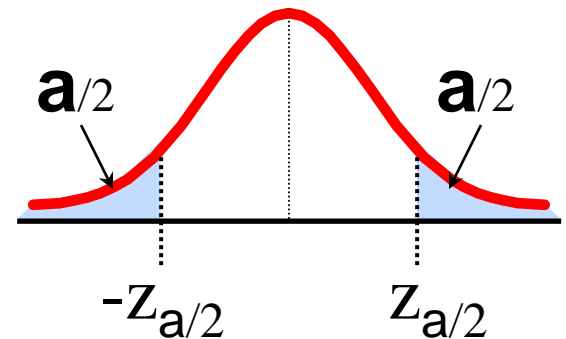
But what is the actual  $p$ -value?

$$P(Z < -1.79) + P(Z > 1.79) = ?$$

Two-tail test:

$$H_0: p_N - p_C = 0$$

$$H_1: p_N - p_C \neq 0$$

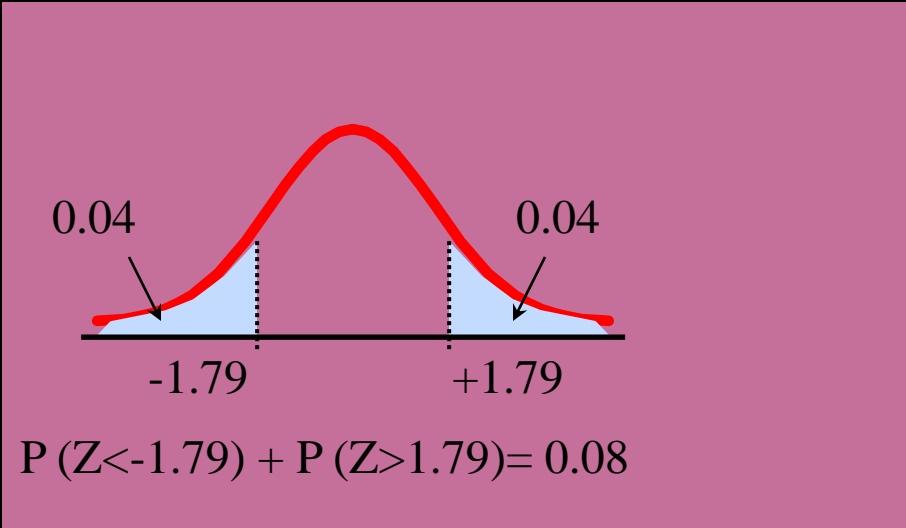


$$Z_{\alpha/2} = 1.96$$

Reject  $H_0$  if  $Z < -Z_{\alpha/2}$   
or  $Z > Z_{\alpha/2}$

Table 1: Table of the Standard Normal Cumulative Distribution Function  $\Phi(z)$

z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
2.4	0.0002	0.0002	0.0002	0.0002	0.0003	0.0003	0.0003	0.0003	0.0003	0.0002
004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0003
006	0.0006	0.0006	0.0006	0.0005	0.0005	0.0005	0.0005	0.0005	0.0005	0.0005
008	0.0008	0.0008	0.0008	0.0008	0.0007	0.0007	0.0007	0.0007	0.0007	0.0007
012	0.0011	0.0011	0.0011	0.0011	0.0010	0.0010	0.0010	0.0010	0.0010	0.0010
016	0.0016	0.0015	0.0015	0.0015	0.0014	0.0014	0.0014	0.0014	0.0014	0.0014
023	0.0022	0.0021	0.0021	0.0021	0.0020	0.0020	0.0020	0.0020	0.0019	0.0019
031	0.0030	0.0029	0.0029	0.0028	0.0027	0.0027	0.0027	0.0027	0.0026	0.0026
041	0.0040	0.0039	0.0039	0.0038	0.0037	0.0037	0.0037	0.0037	0.0036	0.0036
055	0.0054	0.0052	0.0052	0.0051	0.0049	0.0049	0.0049	0.0049	0.0048	0.0048
073	0.0071	0.0069	0.0069	0.0068	0.0066	0.0066	0.0066	0.0066	0.0064	0.0064
096	0.0094	0.0091	0.0091	0.0089	0.0087	0.0087	0.0087	0.0087	0.0084	0.0084
125	0.0122	0.0119	0.0119	0.0116	0.0113	0.0113	0.0113	0.0113	0.0110	0.0110
162	0.0158	0.0154	0.0154	0.0150	0.0146	0.0146	0.0146	0.0146	0.0143	0.0143
207	0.0202	0.0197	0.0197	0.0192	0.0188	0.0188	0.0188	0.0188	0.0183	0.0183
262	0.0256	0.0250	0.0250	0.0244	0.0239	0.0239	0.0239	0.0239	0.0233	0.0233
-1.8	0.0359	0.0351	0.0344	0.0336	0.0329	0.0322	0.0314	0.0307	0.0301	0.0294
-1.7	0.0446	0.0436	0.0427	0.0418	0.0409	0.0401	0.0392	0.0384	0.0375	0.0367
-1.6	0.0548	0.0537	0.0526	0.0516	0.0505	0.0495	0.0485	0.0475	0.0465	0.0455
-1.5	0.0668	0.0655	0.0643	0.0630	0.0618	0.0606	0.0594	0.0582	0.0571	0.0559
-1.4	0.0808	0.0793	0.0778	0.0764	0.0749	0.0735	0.0721	0.0708	0.0694	0.0681
-1.3	0.0968	0.0951	0.0934	0.0918	0.0901	0.0885	0.0869	0.0853	0.0838	0.0823
-1.2	0.1151	0.1131	0.1112	0.1093	0.1075	0.1056	0.1038	0.1020	0.1003	0.0985
-1.1	0.1357	0.1335	0.1314	0.1292	0.1271	0.1251	0.1230	0.1210	0.1190	0.1170
-1.0	0.1587	0.1562	0.1539	0.1515	0.1492	0.1469	0.1446	0.1423	0.1401	0.1379
-0.9	0.1841	0.1814	0.1788	0.1762	0.1736	0.1711	0.1685	0.1660	0.1635	0.1611
-0.8	0.2119	0.2090	0.2061	0.2033	0.2005	0.1977	0.1949	0.1922	0.1894	0.1867
-0.7	0.2420	0.2389	0.2358	0.2327	0.2296	0.2266	0.2236	0.2206	0.2177	0.2148
-0.6	0.2743	0.2709	0.2676	0.2643	0.2611	0.2578	0.2546	0.2514	0.2483	0.2451
-0.5	0.3085	0.3050	0.3015	0.2981	0.2946	0.2912	0.2877	0.2843	0.2810	0.2776



## p-value

- After calculating a test statistic we convert this to a p-value by comparing its value to distribution of test statistic's under the null hypothesis

- Measure of how likely the test statistic value is under the null hypothesis

$p\text{-value} \leq \alpha \Rightarrow \text{Reject } H_0 \text{ at level } \alpha$

$p\text{-value} > \alpha \Rightarrow \text{Do not reject } H_0 \text{ at level } \alpha$

# What is a $p$ -value?

- 'p' stands for probability
  - Tail area probability based on the observed effect
  - Calculated as the probability of an effect as large as or larger than the observed effect (more extreme in the tails of the distribution), assuming null hypothesis is true
- Measures the strength of the evidence against the null hypothesis
  - Smaller  $p$ -values indicate stronger evidence against the null hypothesis





## *Stating the Conclusions of our Results*

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- When the  $p$ -value is small, we **reject** the null hypothesis or, equivalently, we accept the alternative hypothesis.
  - “Small” is defined as a  $p$ -value  $\leq \alpha$ , where  $\alpha$  = acceptable false (+) rate (usually 0.05).
- When the  $p$ -value is not small, we conclude that we cannot reject the null hypothesis or, equivalently, there is not enough evidence to reject the null hypothesis.
  - “Not small” is defined as a  $p$ -value  $> \alpha$ , where  $\alpha$  = acceptable false (+) rate (usually 0.05).

# STATISTICALLY SIGNIFICANT AND NOT STATISTICALLY SIGNIFICANT

- Statistically significant  
Reject  $H_0$

Sample value not compatible with  $H_0$

Sampling variation is an unlikely explanation of discrepancy between  $H_0$  and sample value

- Not statistically significant  
Do not reject  $H_0$

Sample value compatible with  $H_0$

Sampling variation is an likely explanation of discrepancy between  $H_0$  and sample value

# P-values

Trial	Number dead / randomized		Risk Ratio	95% C.I.	P value
	Intravenous nitrate	Control			
Chiche	3/50	8/45	0.33	(0.09,1.13)	0.08
Some evidence against the null hypothesis					
Flaherty	11/56	11/48	0.83	(0.33,2.12)	0.70
Very weak evidence against the null hypothesis...very likely a chance finding					
Lis	5/64	10/76	0.56	(0.19,1.65)	0.29
Jugdutt	24/154	44/156	0.48	(0.28, 0.82)	0.007

Very strong evidence against the null hypothesis...very unlikely to be a chance finding

# Interpreting *P* values

## If the null hypothesis were true...

Trial	Number dead / randomized		Risk Ratio	95% C.I.	P value
	Intravenous nitrate	Control			
Chiche	3/50	8/45	0.33	(0.09,1.13)	0.08
...8 out of 100 such trials would show a risk reduction of 67% or more extreme just by chance					
Flaherty	11/56	11/48	0.83	(0.33,2.12)	0.70
...70 out of 100 such trials would show a risk reduction of 17% or more extreme just by chance...very likely a chance finding					
Lis	5/64	10/76	0.56	(0.19,1.65)	0.29
Jugdutt	24/154	44/156	0.48	(0.28, 0.82)	0.007

Very unlikely to be a chance finding

# Interpreting *P* values

<b>Trial</b>	<b>Intravenous nitrate</b>	<b>Control</b>	<b>Risk ratio</b>	<b>95% confidence interval</b>	<b><i>P</i> value</b>
Chiche	3/50	8/45	0.33	(0.09, 1.13)	0.08
Bussman	4/31	12/29	0.24	(0.08, 0.74)	0.01
Flaherty	11/56	11/48	0.83	(0.33, 2.12)	0.7
Jaffe	4/57	2/57	2.04	(0.39, 10.71)	0.4
Lis	5/64	10/77	0.56	(0.19, 1.65)	0.29
Jugdutt	12/77	44/157	0.48	(0.28, 0.82)	0.007

- Size of the p-value is related to the sample size
- Lis and Jugdutt trials are similar in effect (~ 50% reduction in risk)...but Jugdutt trial has a large sample size

# Interpreting *P* values

Trial	Intravenous nitrate	Control	Risk ratio	95% confidence interval	<i>P</i> value
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- Size of the p-value is related to the effect size or the observed association or difference
- Chiche and Flaherty trials approximately same size, but observed difference greater in the Chiche trial

# P values

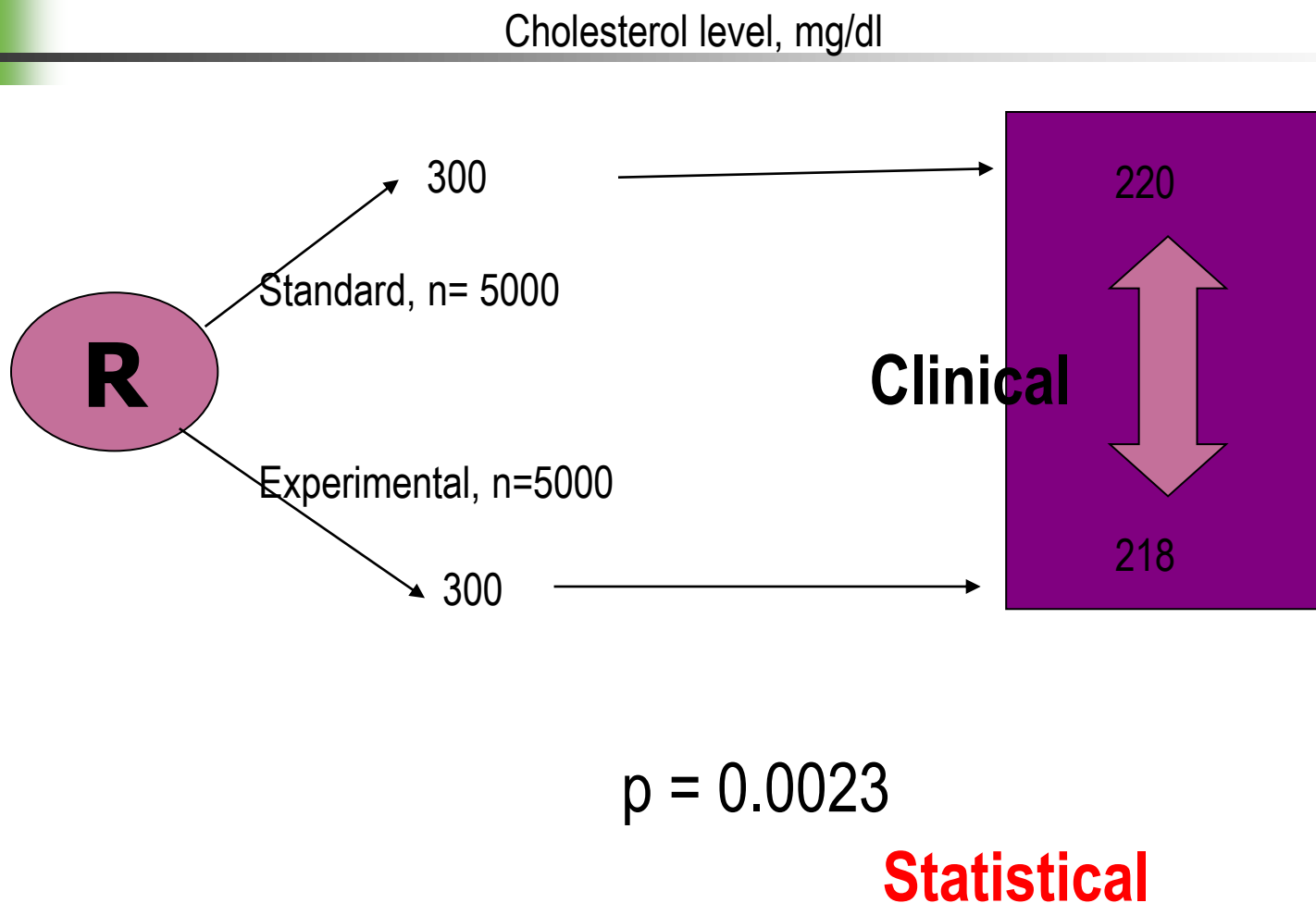
- P values give no indication about the clinical importance of the observed association
- A very large study may result in very small p-value based on a small difference of effect that may not be important when translated into clinical practice
- Therefore, important to look at the effect size and confidence intervals...

Example: If a new antihypertensive therapy reduced the SBP by 1mmHg as compared to standard therapy we are not interested in swapping to the new therapy.

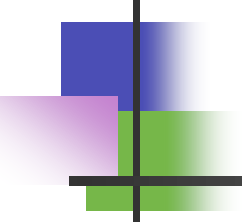
- However, if the decrease was as large as 10 mmHg, then you would be interested in the new therapy.
- Thus, it is important to not only consider whether the difference is statistically significant by the possible magnitude of the difference should also be considered.



# Clinical importance vs. statistical significance



# Clinical importance vs. statistical significance



	Yes	No
Standard	0	10
New	3	7

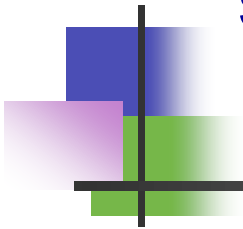
Absolute risk reduction = 30%

← **Clinical**



Fischer exact test:  $p = 0.211$

← **Statistical**

# Reaction of investigator to results of a statistical significance test



## Statistical significance

		Statistical significance	
		Not significant	Significant
Practical importance of observed effect	Not important		Annoyed 
	Important	Very sad 	Elated 