

# Statistical significance using *p*-value

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## Learning Objectives

- (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.

## Investigation

Data **Collection Inferential Statistics Data Presentation Inferential statistics Descriptive Statistics Estimation Hypothesis Measures of Location Univariate analysis Tabulation Testing Measures of Dispersion Diagrams Point estimate Measures of Skewness Multivariate analysis Graphs Interval estimate** & Kurtosis

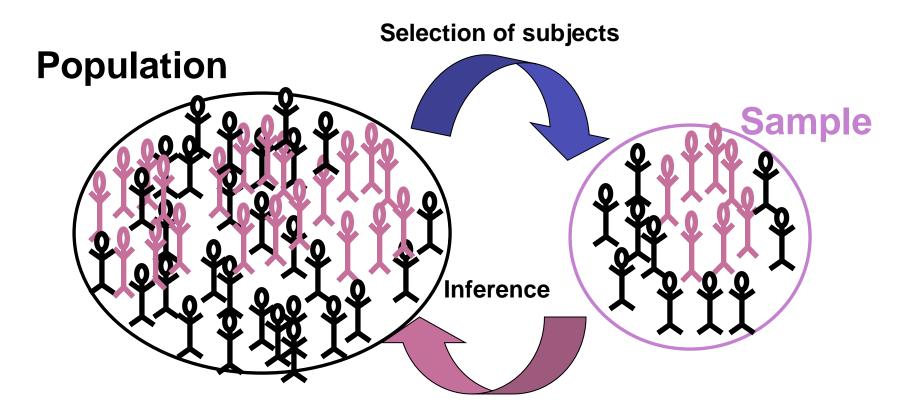
### Why use inferential statistics at all?

Average height of <u>all</u> 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.

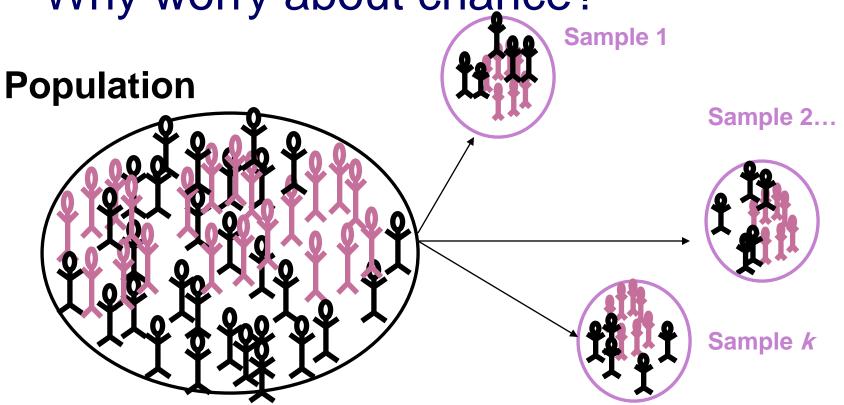
# Is risk factor X associated with disease Y?



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?

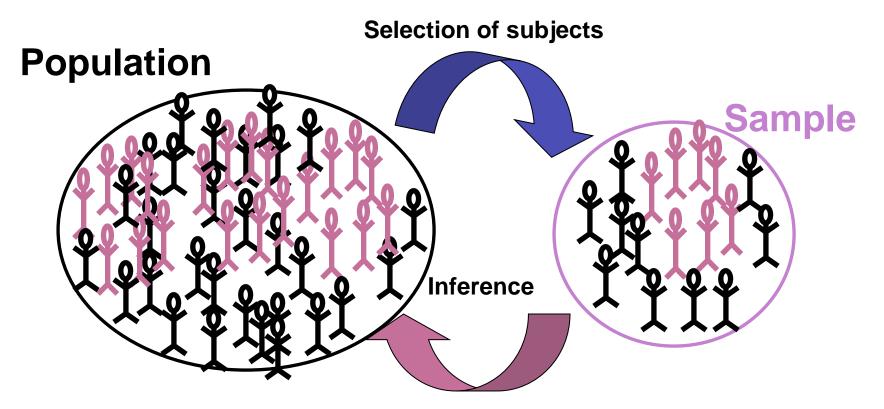
Why worry about chance?



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

#### 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, H0

#### The alternative hypothesis, Ha

:

- 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:

H0: 
$$\mu$$
1-  $\mu$ 2= 0

HA:  $\mu 1 - \mu 2 > 0$  or HA:  $\mu 1 - \mu 2 < 0$ 

Two tailed tests are not directional:

H0:  $\mu$ 1-  $\mu$ 2= 0

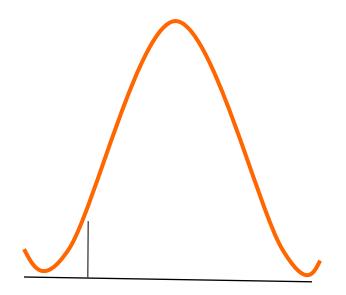
HA:  $\mu 1 - \mu 2 \neq 0$ 



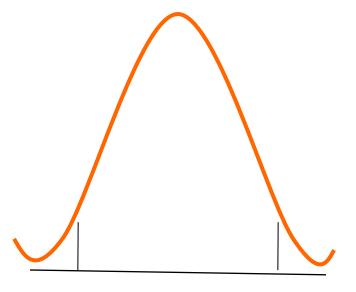
Level of significance, α: Specified before an experiment to define rejection region

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**

- $\alpha$  = Probability of rejecting H<sub>0</sub> when H<sub>0</sub> is true
- α is called significance level of the test
- $\beta$  = Probability of not rejecting H<sub>0</sub> when H<sub>0</sub> is false
- 1-β is called statistical power of the test

## Diagnosis and statistical reasoning

#### **Disease status**

Present Absent

#### Test result

+ve True +ve False +ve

(sensitivity)

-ve False –ve True -ve

(Specificity)

Significance Difference is			
	Present	Absent	
	(Ho not true)	(Ho is true)	
Test result Reject Ho	No error 1-β	Type I err. α	
Accept Ho	Type II err.	No error	

 $1-\alpha$ 

 $\alpha$ : significance level

β

1-β: power

## Significance testing

Subjects with Acute MI

Mortality
IV nitrate

P<sub>N</sub>

Mortality
No nitrate

P<sub>C</sub>

- 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<sub>o</sub>)

- There is no association between the independent and dependent/outcome variables
  - Formal basis for hypothesis testing

In the example, H<sub>o</sub>: "The administration of IV nitrate has no effect on mortality in MI patients" or P<sub>N</sub> - P<sub>C</sub> = 0

### Obtaining P values

	Number dead / randomized				
Trial	Intravenous	Control	Risk Ratio	95% C.I. P value	
	nitrate		How do we get this <i>p</i> -value?		
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	

### Example of significance testing

In the Chiche trial:

$$p_N = 3/50 = 0.06; p_C = 8/45 = 0.178$$

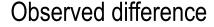
Null hypothesis:

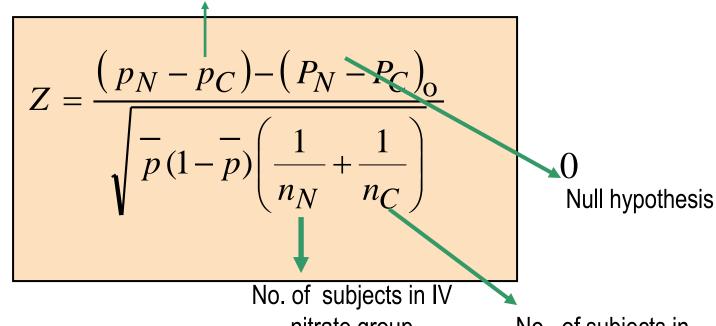
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$$H_0$$
:  $\rho_N - \rho_C = 0$  or  $\rho_N = \rho_C$ 

- Statistical test:
  - Two-sample proportion

#### Test statistic for Two Population Proportions

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



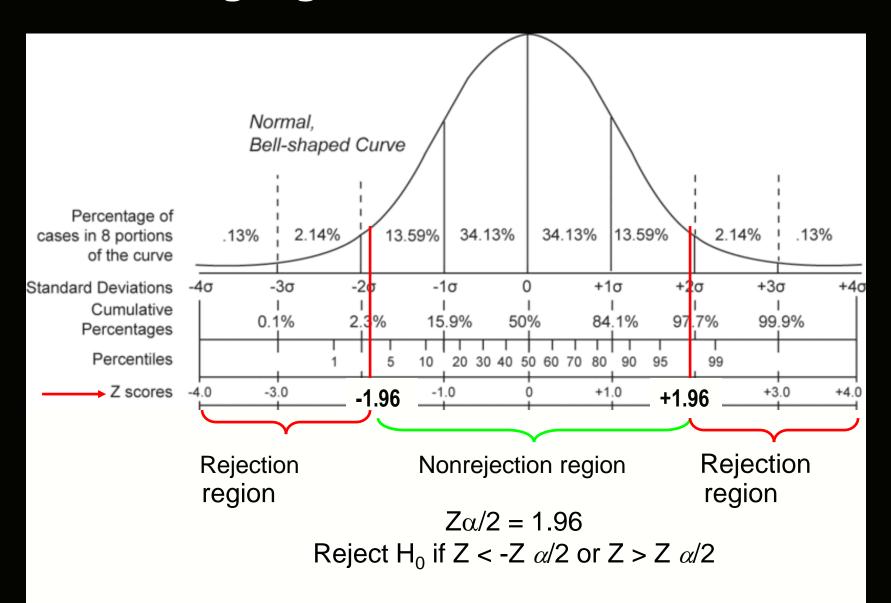


nitrate group

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



## **Two Population Proportions**

(continued)

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

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<sub>1</sub> – p<sub>2</sub>

Two Population Proportions, Independent Samples

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

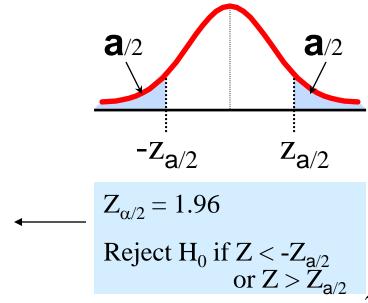
Two-tail test:

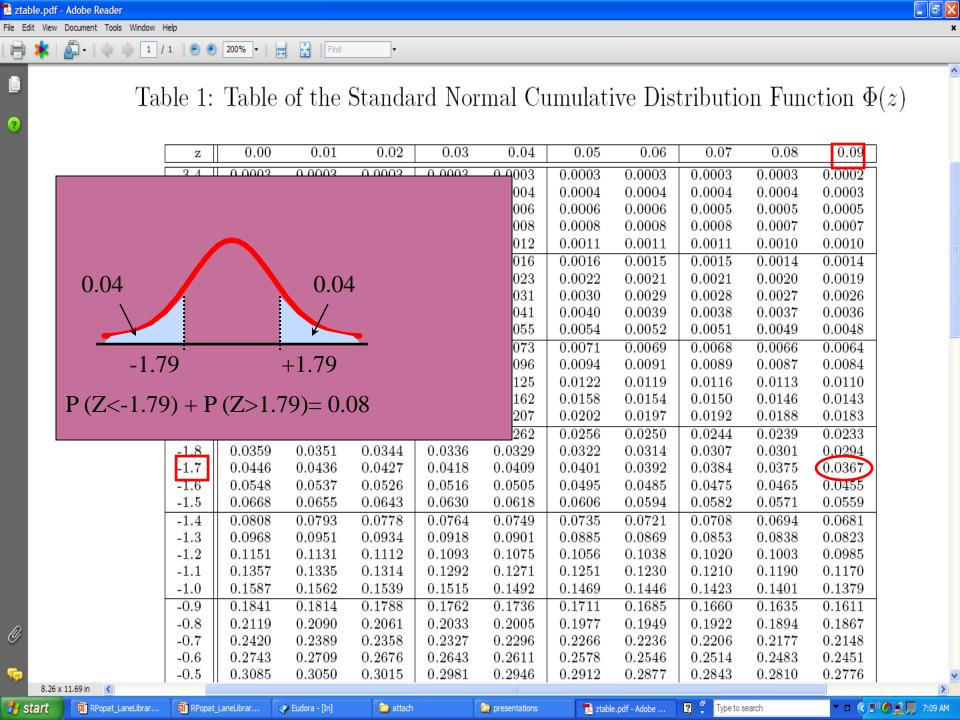
$$H_0: p_N - p_C = 0$$
  
 $H_1: p_N - p_C \neq 0$ 

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

But what is the actual *p*-value?

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





#### 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-value  $\leq \alpha \Rightarrow \text{Reject H}_0$  at level  $\alpha$  p-value  $> \alpha \Rightarrow \text{Do not reject H}_0$  at level  $\alpha$

#### Stating the Conclusions of our Results

- 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).

#### 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

## STATISTICALLY SIGNIFICANT AND NOT STATISTICALLY SINGIFICANT

 Statistically significant
 Reject Ho

Sample value not compatible with Ho

Sampling variation is an unlikely explanation of discrepancy between Ho and sample value  Not statistically significant
 Do not reject Ho

Sample value compatible with Ho

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

#### *P*-values

	Number dead / randomized			
Trial	Intravenous nitrate	Control	Risk Ratio	95% C.I. P value
Chiche	3/50	8/45	0.33	(0.09,1.13) 0.08
Some evidend	ce against the r	null hypothes	sis	
Flaherty	11/56	11/48	0.83	(0.33,2.12) 0.70
Very weak evi	dence against	the null hype	othesisver	y likely a chance
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...

	Number dead / rai	ndomized			
Trial	Intravenous nitrate	Control	Risk Ratio	95% C.I.	P value
Chiche	3/50	8/45	0.33	(0.09,1.13)	0.08
8 out of 100 extreme just b	such trials woul	d show a ris	sk reductior	n of 67% or m	ore
Flaherty	11/56	11/48	0.83	(0.33,2.12)	0.70
	0 such trials wou			on of 17% or r	more
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					

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### Interpreting P values

Trial	Intravenous nitrate	Control	Risk ratio	95% confidence interval	P value
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	P value
Chiche	3/50	8/45	0.33	(0.09, 1.13)	0.08
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Jugdutt	12/77	44/157	0.48	(0.28, 0.82)	0.007

- 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

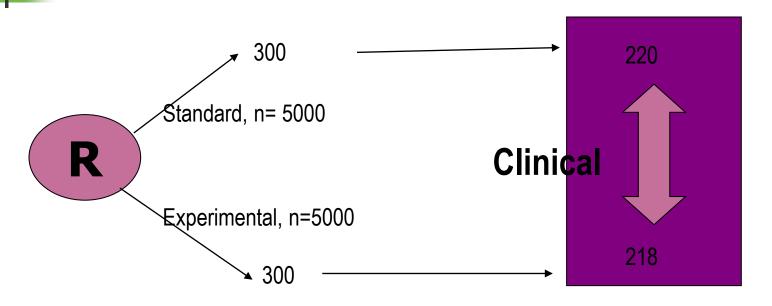
- P values give no indication about the clinical importance of the observed association
- A very large study may result in very small pvalue based on a small difference of effect that may not be important when translated into clinical practice
- Therefore, important to look at the effect size and confidence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

Standard 0 10

New 3 7

Absolute risk reduction = 30%



Fischer exact test: p = 0.211



## Reaction of investigator to results of a statistical significance test

#### Statistical significance

Practical importance of observed effect

	Not significant	Significant
Not important		Annoyed <u></u>
Important	Very sad	Elated 😃