



# Hypothesis Testing

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# Overview

## Outline

- Review
- Hypothesis testing
- Examples

## Learning objectives

- to be able to state a good hypothesis
- to be aware of steps for hypothesis testing
- to be able to make decisions based on hypothesis tests



# Review

## Terminology

- **population**

group of individuals or objects that we would like to study

- **sample**

subset of a population



# Review

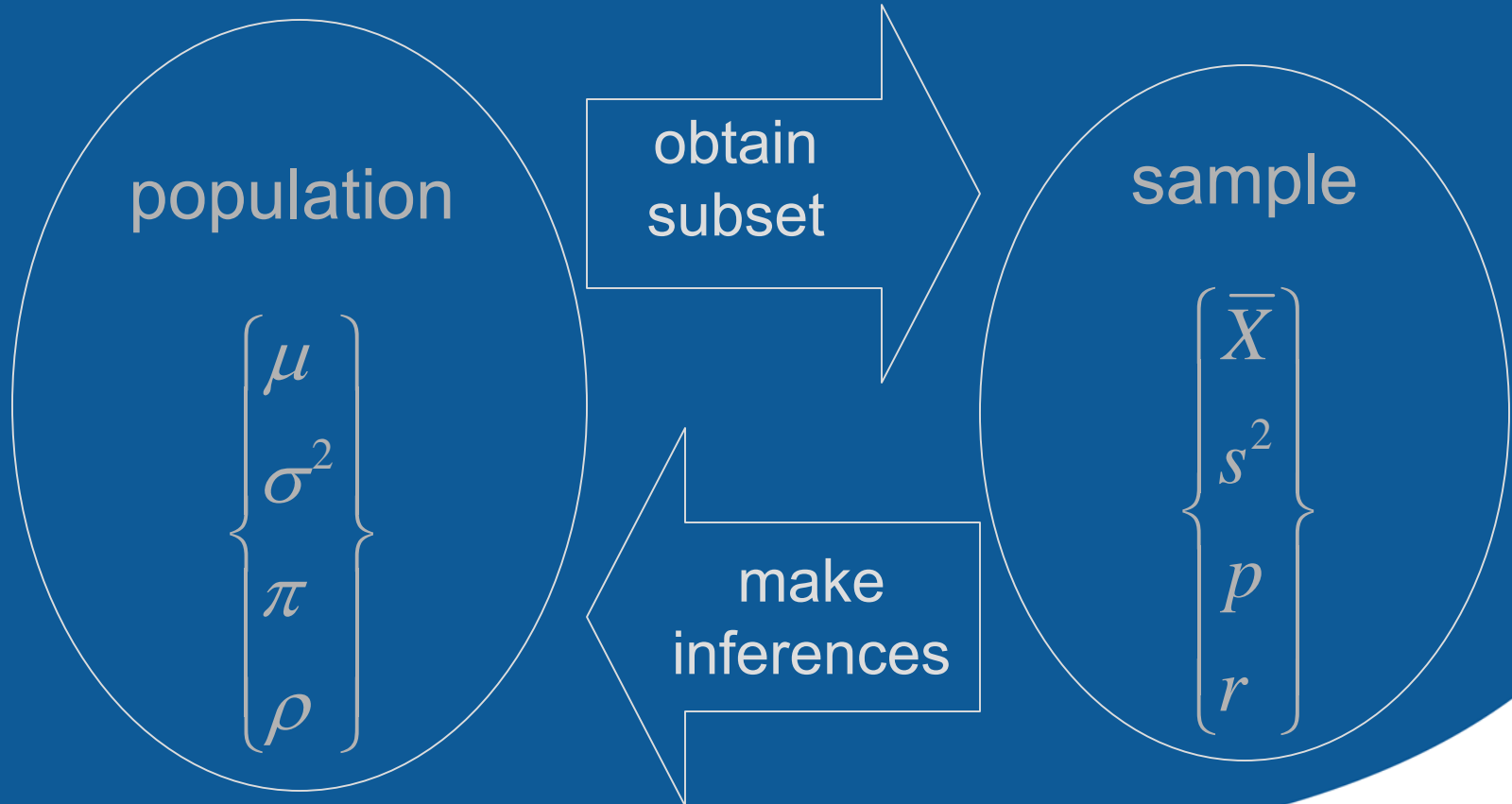
## Terminology

- **population parameter**  
an unknown quantity that describes a population
- **sample statistic**  
an estimate of the population parameter
- **statistical inference**  
process of drawing conclusions about a population based on observations in a sample



# Review

## Framework for statistical inference



# Introduction

## Motivation of hypothesis testing

- to help make decisions
- to determine if chance alone explains our observations



# Hypothesis testing

## Hypothesis testing

- process of making decisions based on the likelihood of the data

## Main components

- develop hypothesis
- test hypothesis
- make decision



# Develop the hypothesis

## Research question

- define study population
- specify question of interest

## Statistical question

- translate research question into statistical question
- identify population parameter
- state hypothesis



# Develop the hypothesis

## State hypothesis

- **null hypothesis,  $H_0$**   
typically states that there is no relationship between the response and explanatory variable(s)
- **alternative hypothesis,  $H_A$**   
typically states that there is a relationship between the response and explanatory variable(s)



# Develop the hypothesis

## Components to think about

- population(s)
- population parameter
- directional or non-directional



# Develop the hypothesis

## Set significance level, $\alpha$

- cutoff point that determines what decision will be made concerning the null hypothesis
- $\alpha$  is between 0 and 1 (eg. 0.10, 0.05, 0.01)
- the probability with which the researcher is willing to reject  $H_0$  when it is actually true



# Develop the hypothesis

## Expressions for the significance level, $\alpha$

- alpha level ( $\alpha$ ) was fixed at 0.05
- $H_0$  will be rejected if  $p < 0.05$
- statistical significance was set at the 5% level
- differences between groups were examined at the 0.05 significance level



# Develop the hypothesis

## Check your understanding

- state a “good” hypothesis



# Test hypothesis

## Data collection

- collect and summarize sample data
- researcher's decision to reject  $H_0$  is made after observing sample data
- evidence to reject  $H_0$  is based on sample data that is converted into a numeric value (ie. test statistic or p-value)



# Test hypothesis

## Test statistics

- quantify evidence to reject or not reject  $H_0$
- have some distribution based on theory



# Test hypothesis

## Test statistics

- most test statistics,  $t_{obs}$ , have the following form:

$$t_{obs} = \frac{\text{sample statistic} - \text{population parameter under } H_0}{\text{standard error of sample statistic}}$$

- measure the difference between the value from the data and the hypothesized value, and scales that value by the variation in such sample statistics based on the data (i.e. the standard error)



# Test hypothesis

## p-value

- quantify evidence to reject or not reject  $H_0$
- probability of observed data assuming the null hypothesis is true
- probability of observing a test statistic that is as extreme or more extreme than currently observed assuming the null hypothesis is true



# Test hypothesis

## Check your understanding

- name a test statistic



# Make decision

## critical value

- value of the statistic marking the boundary between the acceptance region and rejection region

## significance level, $\alpha$

- predetermined probability with which the researcher is willing to reject  $H_0$  when it is actually true



# Make decision

## Decision criterion

- compare the test statistic with the critical value  
→ test statistics greater (in absolute value) than the critical value indicate reject  $H_0$
- compare the sample-based p-value with the significance level,  $\alpha$   
→ p-values less than  $\alpha$  indicate reject  $H_0$



# Make decision

## Expressions to reject $H_0$

- $H_0$  was rejected
- statistically significant
- p-value is less than 0.05
- there is significant evidence to reject  $H_0$



# Make decision

## Expressions to fail to reject $H_0$

- $H_0$  was not rejected
- not statistically significant
- p-value is greater than 0.05
- no significant evidence to reject  $H_0$



# Make decision

## Interpret the decision

- translate decision into context of research
- does statistical significance imply clinical significance?
- does statistical non-significance imply clinical non-significance?



# Make decision

## Check your understanding

- how are decisions are made?



# Example: scenario

Suppose that you would like to compare the effect of a newly developed drug (drug A) and a current drug (drug B) on systolic blood pressure (sbp).

How should we proceed?



# Example: develop hypothesis

## What is the hypothesis?

$H_0$ : the mean sbp among patients on drug A is the same as the mean sbp among patients on drug B  
ie.  $\mu_A = \mu_B$  or  $\mu_A - \mu_B = 0$

$H_A$ : the mean sbp among patients on drug A is different from the mean sbp of patients on drug B  
ie.  $\mu_A \neq \mu_B$  or  $\mu_A - \mu_B \neq 0$



# Example: develop hypothesis

significance level,  $\alpha$

Q: How often would you be comfortable with rejecting  $H_0$  when there is no difference?

A: 1 out of 20 times  
→  $\alpha = 0.05$



# Example: test hypothesis

## What does the data show?

- we took a random sample of 35 patients on drug A and 35 patients on drug B.
- we observed an average sbp of 107 mmHg and 125 mmHg among drug A and drug B patients, respectively.
- we observed an sd of 19 and 20 mmHg, respectively.

## How will we test our hypothesis?

independent two-sample t-test  
with two-sided alternative



# Example: test hypothesis

What is the test statistic?

$$t_{\text{obs}} = \frac{\bar{x}_B - \bar{x}_A}{s_p \sqrt{\frac{1}{n_B} + \frac{1}{n_A}}} = \frac{125 - 107}{19.5 \sqrt{\frac{1}{35} + \frac{1}{35}}} \approx 3.9$$

$$s_p = \sqrt{\frac{(n_A - 1)s_A^2 + (n_B - 1)s_B^2}{n_A + n_B - 2}} = \sqrt{\frac{(35 - 1)(19^2) + (35 - 1)(20^2)}{35 + 35 - 2}}$$

$\approx 19.5$  mmHg



# Example: make decision

## Decision criterion

Find critical value:

$$t_{\alpha/2, n_1+n_2-2} = 2.0$$

Compare test statistic with critical value:

$$t_{\text{obs}} = 3.9 > 2.0 = t_{\alpha/2, n_1+n_2-2}$$

Make decision:

➔ reject  $H_0$



# Example: make decision

## Decision criterion

Find p-value:

$$p\text{-value} = 0.0001$$

Compare p-value with  $\alpha$ :

$$p\text{-value} = 0.0001 < 0.05 = \alpha$$

Make decision:

 reject  $H_0$



# Example: make decision

## Decision

There is evidence to support that there is a significant difference in mean sbp between patients on drug A and patients on drug B ( $p=0.0001$ ).



# Example: confidence interval

95% CI for difference between mean sbp:

$$(\bar{x}_B - \bar{x}_A) \pm t_{\alpha/2, n_A+n_B-2} s_p \sqrt{\frac{1}{n_A} + \frac{1}{n_B}}$$

$$\Rightarrow (125 - 107) \pm (2.0)(4.7)$$

$$\Rightarrow 18 \pm 9.3$$

$$\Rightarrow (8.7, 27.3) \text{ mmHg}$$



# Example: confidence interval

## Interpretation

The mean sbp for patients on drug B is on average 18 (95% CI 9-27) mmHg higher than the mean sbp for patients on drug A.



# Relation to confidence intervals

## Hypothesis testing and confidence intervals

- hypothesis testing and confidence intervals are related
- $H_0$  is not rejected at the  $\alpha$  significance level if the  $100(1 - \alpha)\%$  CI captures the parameter value specified under  $H_0$
- $H_0$  is rejected at the  $\alpha$  significance level if the  $100(1 - \alpha)\%$  CI does not capture the parameter value specified under  $H_0$



# Fallacies

- statistical significance implies clinical significance
- small p-values imply large effects
- not rejecting  $H_0$  implies that it is true
- p-value is the probability that  $H_0$  is true



# Examples from literature

## Check your understanding

- find some examples of hypotheses
- find some examples of how decisions were made
- find some examples of decisions



# Errors in hypothesis testing

- **type I error,  $\alpha$**   
chance of mistakenly rejecting  $H_0$
- **type II error,  $\beta$**   
chance of mistakenly accepting  $H_0$



# Errors in hypothesis testing

## Truth table

	State of Reality	
Decision	$H_0$ true	$H_0$ false
Reject $H_0$	type I error $\alpha$	correct decision $1-\beta$
Accept $H_0$	correct decision $1-\alpha$	type II error $\beta$



# Follow-up

## Thoughts

- how do we control the chance of error?
  - how do we know if we were able to detect a clinically significant result?
- evaluate power and sample size before beginning study



# References

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