



Basics of Health Intelligent Data Analysis
PhD Programme in Health Data Science

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- 1. The coin example
- 2. Hypothesis testing for the mean
- 3. Types of error in hypothesis testing









- With confidence intervals we can infer about a parameter in the population based on an estimate of that parameter.
- Hypothesis tests are based on another different (but related) approach.
- The idea now is to measure how much the results observed in the sample are compatible with a hypothesis about the population.



#### 1. Define the hypothesis

H0 = null hypothesis - no effect on population

- 2. Set the significance level (alpha) usually 0.05
- **3. Get the test statistic with the sample data and obtain the p-value:** probability of obtaining the result we obtained or even more extreme, assuming H0 is true.

#### **5.** Interpret the p-value:

if p < alpha, we have enough evidence to reject H0 if  $p \ge alpha$ , we do not have enough evidence to reject H0



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N Engl J Med 2003; 349:1000-1002.

Results: Subjects on the low-carbohydrate diet had lost more weight than subjects on the conventional diet at 3 months (mean [ $\pm$ SD],  $-6.8\pm5.0$  vs.  $-2.7\pm3.7$  percent of body weight; P=0.001) and 6 months ( $-7.0\pm6.5$  vs.  $-3.2\pm5.6$  percent of body weight, P=0.02), but the difference at 12 months was not significant ( $-4.4\pm6.7$  vs.  $-2.5\pm6.3$  percent of body weight, P=0.26).



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The classic example: Toss a coin 10 times in the air; observe heads or tails.

Suppose we get 5 heads (5 tails). Is the coin balanced?

Suppose we get 7 heads (3 tails). Is the coin balanced?

Suppose we get 9 heads (1 tails). Is the coin balanced?

Suppose we get 10 heads (0 tails). Is the coin balanced?





Being the coin balanced, we expected 5 heads and 5 tails (null hypothesis).

Could the results we found (6 heads and 4 tails; 7 heads and 3 tails;...; 10 heads and 0 tails) be due to chance?

- Yes, it can always be due to chance

But to what extent do we accept chance as justification for the obtained result?





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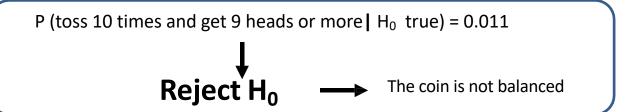
Suppose we get 5 heads (5 tails). Is the coin balanced? Probability 0.25
Suppose we get 7 heads (3 tails). Is the coin balanced? Probability 0.12
Suppose we get 9 heads (1 tails). Is the coin balanced? Probability 0.01
Suppose we get 10 heads (0 tails). Is the coin balanced? Probability 0.001





The null hypothesis is typically the opposite of the hypothesis we want to investigate (no effect)

H0: coin is balanced or heads ratio = 0.5





In the coin example it is more or less intuitive to assess the extent to which we accept chance as justification for the result obtained.

But in other cases it becomes more difficult.



- In the clinical trial comparing the two diets, it was observed that the average weight reduction in one of them was 6.8kg and in the other 2.7kg.
- Is this result very or little compatible with the null hypothesis that diets are identical?

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H<sub>0</sub>: 
$$\mu_1 = \mu_0$$
 or H<sub>0</sub>:  $\mu_1 - \mu_0 = 0$ 

$$X_1 = -6.8 \text{ e } X_2 = -2.7$$

Is there evidence against the null hypothesis?

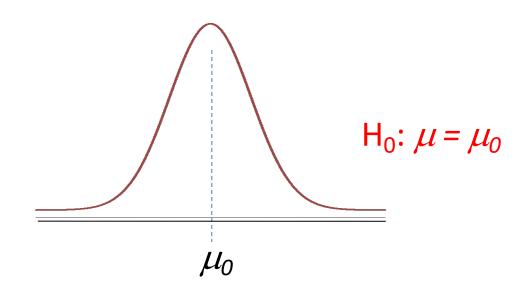


#### By the **Central Limit Theorem**,

the distribution of the sample means of size n tends to a normal distribution with mean equal to population mean and standard deviation equal to standard error

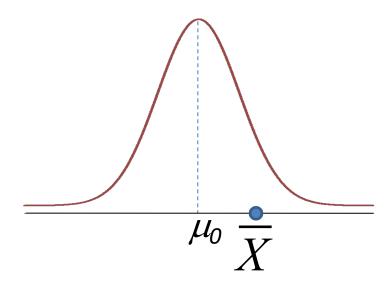


• If null hypothesis is true



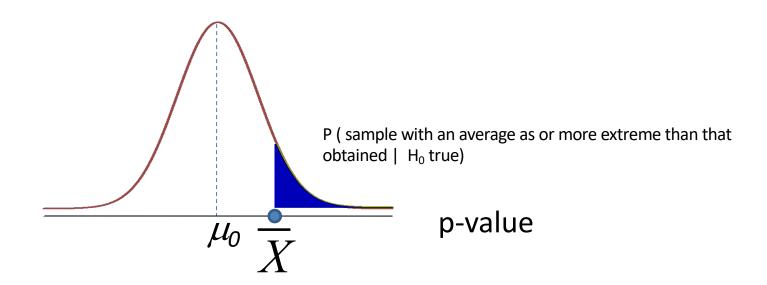


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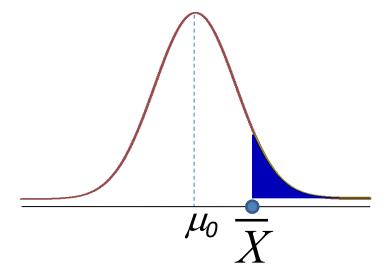




**The p-value** is the probability of obtaining a result as or more extreme than that observed in the sample, since (assuming) the null hypothesis is true.



The observed sample mean seems compatible with the null hypothesis.

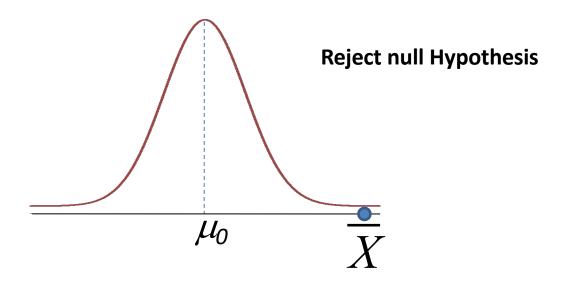






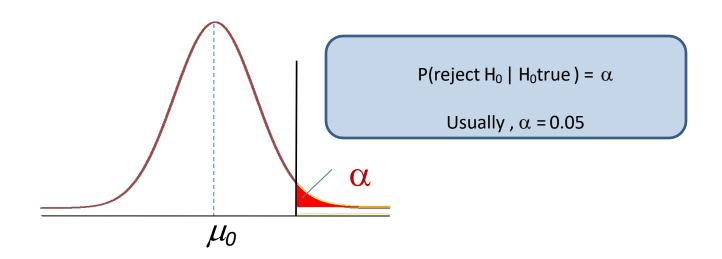


A sample with this mean was hardly obtained from a population where the null hypothesis is true.





But from what value do we decide we should reject the null hypothesis?

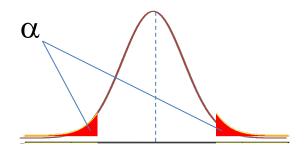




The previous example corresponds to "one-tail" or "one-side" tests where the alternative hypothesis (H1) corresponds to

$$H_1: \mu_1 > \mu_0$$

It is, however, more common for the alternative hypothesis (H<sub>1</sub>) to be H<sub>1</sub>:  $\mu_1 \neq \mu_0$  corresponding to a **two-sided test** 







	H <sub>0</sub> true	H <sub>o</sub> false
Accept H <sub>0</sub>	Without error	Type II error (β)
Reject H <sub>0</sub>	Type I error (α)	Without error(1-β)

**test power** =  $1 - \text{beta} = \text{probability of rejecting H}_0$  when false

that is,

the ability of the test to detect differences when they exist



# Hypothesis test - error

 $\alpha$  is defined by us when we set the level of significance

To have a higher test power  $(1-\beta)$  we need:

- larger sample size
- smaller variability of observations
- higher effect
- higher  $\alpha$