



11. Introduction to Hypothesis Testing. Large-Sample Test about a Mean



Hypothesis testing

The purpose of hypothesis testing is to determine whether there is enough statistical evidence in favor of a certain belief about a parameter.

Examples:

Is there any statistical evidence in a random sample of potential customers that supports the hypothesis that more than 50% of the potential customers will purchase new products?

Is a new drug effective in curing a certain disease? A sample of patients is randomly selected. Half of them are given the drug where half are given a placebo (sugar pill). The improvement in the patients' conditions is then measured and compared.



Concept of hypothesis testing: two hypotheses

There are two hypotheses about a population parameter(s).

- The *null hypothesis*, denoted by H_0 , is a claim about a population parameter that is *initially* assumed to be true.
- The *alternative (research) hypothesis*, H_a , is a competitive claim about the same parameter.

The null hypothesis will be rejected in favor of the alternative if a sample provides a strong evidence that H_0 is false. If the sample do not suggest such evidence, H_0 will not be rejected.



Concept of hypothesis testing: two possible conclusions

Thus we can have two possible conclusions:

- Reject H_0 , accept H_a
- Fail to reject H_0 , reject H_a



Concept of hypothesis testing: setup

The form of the null hypothesis is

$$H_0: \textit{population parameter} = \textit{test value}$$

Here the *test value* is a specific number determined by the problem context.

For the alternative hypothesis we can have *one* of the following three possibilities:

$$H_a: \textit{population parameter} > \textit{test value}$$

$$H_a: \textit{population parameter} < \textit{test value}$$

$$H_a: \textit{population parameter} \neq \textit{test value}$$

Which one will be used depends on the problem.



Concept of hypothesis testing: type I and II errors

Two types of errors are possible when the decision whether to reject the null hypothesis is made.

		The actual state	
		H_0 is true	H_a is true
Our conclusion	Reject H_a	OK	Type II error
	Accept H_a	Type I error	OK

- Probability of Type I error is denoted by α and called the *significance level*.
- Probability of Type II error is denoted by β .



Concept of hypothesis testing: type I and II errors

Two errors are *inversely* related: if we try to reduce one error the other one will increase. Therefore, we have to balance the consequences of Type I and Type II errors.

The accepted practice is to employ the largest α that is tolerable for the problem, and then we try to do our best in minimizing β . The significance level α is always under our control.



Concept of hypothesis testing: test statistic

Once a sample is collected, the typical testing goes like this.

1. Assume the null hypothesis is true.
2. Build a *test statistic* (based on the sample) related to the parameter of interest.
3. Pose the following question. How probable is it to obtain a test statistic value at least as extreme as the observed test statistic if we assume the null hypothesis is true?
4. Make a conclusion based on your answer to that question.



Concept of hypothesis testing: making conclusion

There are two ways how we can answer the question in step 3.

- ***The rejection region approach.*** Depending on (1) the type of the alternative, (2) distribution of the test statistic under H_0 , and (3) chosen significance level α we can construct a region that corresponds to extreme values of test statistics. If our test statistic is in that region we reject H_0 , and accept H_a .
- ***The p -value approach.*** Again depending on (1) the type of the alternative, (2) distribution of the test statistic under H_0 , and (3) chosen significance level α we calculate the probability (called p -value) of obtaining a test statistic value at least as extreme as our test statistic, assuming that H_0 is true. If this probability, or p -value, is less than α , then we reject H_0 , and accept H_a .



Testing population mean: z -test about μ

The z -test about μ is employed when one wants to check a claim about the population mean, and the sample size n is large, typically, $n \geq 30$.



z-test about μ : setup

The null hypothesis:

$$**$H_0: \mu = \mu_0$**$$

Possible alternatives:

$$**$H_a: \mu > \mu_0$**$$

$$**$H_a: \mu < \mu_0$**$$

$$**$H_a: \mu \neq \mu_0$**$$

Which type of alternative and what test value μ_0 will be used is determined by the problem context.

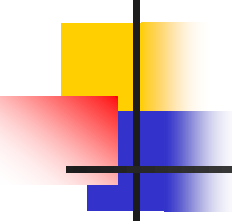


z-test about μ : test statistic

To test the research hypothesis we use the following test statistic

$$z = \frac{\bar{X} - \mu_0}{s/\sqrt{n}}$$

If H_0 is true then by the central limit theorem and the law of large numbers random variable z has approximately standard normal distribution. We use this fact to make our judgment whether the sample is consistent with the null or the alternative.



z-test about μ : rejection region

Three possible expressions for the rejection region:

- **If we test $H_a: \mu > \mu_0$, then $RR = [z_\alpha, +\infty]$**
- **If we test $H_a: \mu < \mu_0$, then $RR = [-\infty, -z_\alpha]$**
- **If we test $H_a: \mu \neq \mu_0$, then $RR = [-\infty, -z_{\alpha/2}] \cup [z_{\alpha/2}, +\infty]$**

The rejection region tells us which values of the test statistic z are not consistent with the null hypothesis H_0 when it is tested versus a given alternative H_a .



z-test about μ : conclusion based on rejection region

The decision depends on the relationship between the rejection region and the test statistic. The universal rule is

- If the test statistic falls *inside* the rejection region we *accept* H_a , the research claim.
- If the test statistic falls *outside* the rejection region we *reject* H_a .



z-test about μ : p -value method

Instead of the rejection region approach we can calculate p -value. There are three formulas depending on a type of the alternative.

- If we test $H_a: \mu > \mu_0$, then $p\text{-value} = P(Z > z)$
- If we test $H_a: \mu < \mu_0$, then $p\text{-value} = P(Z < z)$
- If we test $H_a: \mu \neq \mu_0$, then $p\text{-value} = 2 \times P(Z > |z|)$

p -value is the probability of obtaining a test statistic value at least as extreme as our observed test statistic, assuming that H_0 is true. If this probability is small it means that our observed test statistic is too extreme to be consistent with H_0 .



z-test about μ : conclusion based on p -value

The universal rule is

- If $p\text{-value} < \alpha$, then we *accept* H_a .
- If $p\text{-value} > \alpha$, then we *reject* H_a .



Example

Example 1. In a nationwide opinion poll based on a random sample of 240 people, one question is: “How do you rate the ethics of business executives of large companies?” A rating of 3 means “no better or worse than most people”, a rating of 1 is “much better than most people”, and 5 is “much worse than most people”. The sample mean rating is 3.1 and the standard deviation is 0.9. Can we infer at significance level 5% that respondents rated the ethics of executives at different level in comparison to most people?



1. Setup

Let μ be the true (nationwide) mean rating of business executives of large companies.

$$H_0: \mu = 3$$

$$H_a: \mu \neq 3$$



2. Test statistic

Test statistic is given by

$$z = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} = \frac{3.1 - 3}{.9/\sqrt{240}} \approx 1.72$$



3. Rejection region

The rejection region is given by

$$\begin{aligned} RR &= [-\infty, -z_{\alpha/2}] \cup [z_{\alpha/2}, +\infty] \\ &= [-\infty, -z_{.025}] \cup [z_{.025}, +\infty] \\ &= [-\infty, -\mathbf{1.96}] \cup [\mathbf{1.96}, +\infty] \end{aligned}$$



4. Conclusion

Since the test statistic falls outside the rejection region we reject H_a , that is, there is no difference in ratings between executives and most people.



3` . p -value

The p -value of the test is

$$p\text{-value} = 2 \times P(Z > 1.72) = 2 \times .0427 = .0854$$



4. Conclusion based on p -value

Since $p\text{-value} = .0854 > \alpha = .05$, we reject H_0 , that is, there is a difference in ratings between executives and most people.

An important remark: rejection region approach and p -value approach are mathematically equivalent.



z-test about μ : exercise

***Exercise 1.* According to the U. S. Department of Commerce, the average price for a new home topped \$200,000 for the first time in 1999. In November 1999, the average new-home price was \$209,700 (Wall Street Journal Interactive Edition, Jan. 7, 2000). The prices of a random sample of 32 new homes sold in November 2000 yielded $\bar{X} = 216981$ and $s = 19805$.**

- a. What are the appropriate null and alternative hypotheses to test whether the mean price of a new home in November 2000 exceeds \$209,700?**
- b. Compute and interpret the p-value of the test. Do the data provide sufficient evidence to conclude that the mean new-home price in November 2000 exceeded the reported mean price of November 1999?**