Hypothesis Testing

Basics and Tests

Hypothesis Testing

- Introduction to Hypothesis
- Type I and Type II Errors
- Test of Hypothesis



Example 1:

Pepsico India sells a soft drink in a can container having 250ml of the drink.

Since this information is on the label, we assume it to be true.

But is it really true?





Example 1:

Pepsico India sells a soft drink in a can container having 250ml of the drink.

As a customer:

We want the volume to be at least 250ml. Quantity of Drink >= 250ml

As a manufacturer:

We want the volume of the drink to be exactly 250ml. No more, no less Quantity of Drink = 250ml





So we collected 100 Pepsi cans from all over India (to randomize the sample)

Then we measured the quantity of soft drink in each can in the sample and find the mean quantity for all 100 cans.

Using those sample means or measurements, we can TEST THE ASSUMPTION (i.e. the STATUS QUO)





Example 2:

An auto manufacturer had designed a new engine which claims to reduce the fuel consumption. It claims that the new engine makes more efficient use of fuel and performs better than the old engine which used to run at 30 miles per gallon.

Company now need to run some tests to look for statistical evidence to support the **claim** that the new engine offers better fuel efficiency.





Company Claim:

Fuel Efficiency > 30

But it cold be well above or below 30 mpg.

The manufacturer is making a claim it wishes to test.

It is not testing an assumption (i.e. Status Quo)



Notice Assumption Vs. Claim



When trying to formulate a statistical hypothesis, think about the following:

Are we testing an assumption (status quo) that already exists? (Pepsi Can) Or

Are we testing a claim (assertion) which may be true, if not then truth is something else? (New Engine)

Say, if the manufacturer now tested the engine and fix it in a car and rolled it out in the market saying the car performs 35 mpg; and now we want to test it, we will be testing an assumption here and not a claim.



Null Hypothesis:

The null hypothesis is the initial position. It is the status-quo position. It is the position that is rejected or fails to be rejected. It is the position that needs to be validated. It is the position that needs to be tested.

e.g.: Pepsi Can contains 250 ml of soft drink

Alternate Hypothesis:

The alternate hypothesis is the contrary position to NULL hypothesis. If there are statistically significant evidences that suggest that the alternate hypothesis is valid, then the NULL hypothesis is rejected.

e.g.: Pepsi Can does not contain 250 ml of soft drink (otherwise of NULL)

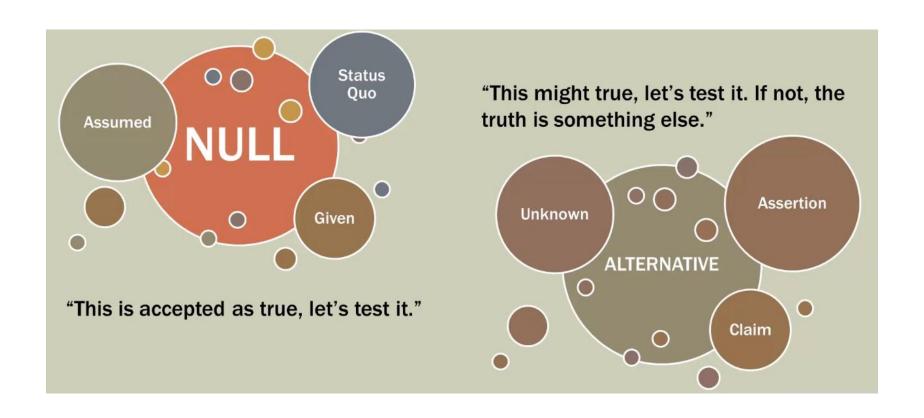


By Definition, null and alternate hypothesis are opposite; mutually exclusive. They both cannot be true.

The null is either rejected or not. Only if the null is rejected can we move with the alternate hypothesis.

Null and alternate hypothesis depends on the question we are trying to ask. A person can start with either the null hypothesis or the alternate hypothesis, and then form the other as a complement to the first.







Hypothesis testing can be used to determine whether a statement about the value of a population parameter should or should not be rejected.

<u>Null hypothesis:</u> denoted by H_0 , is a tentative assumption about a population parameter.

<u>Alternate hypothesis:</u> denoted by $\mathbf{H_a}$, is the opposite of what is stated in the null hypothesis.

e.g. Let's say a court room trial:

H₀: The accused is innocent

H_a: The accused is guilty



H_0	H_a
Assumption, status quo, nothing new	Rejection of an assumption
Assumed to be "true"; a given.	Rejection of an assumption or the given.
Negation of the research question	Research question to be "proven"
Always contains an equality $(=, \leq, \geq)$	Does not contain equality $(\neq, <, >)$

Based on the last property, we can derive the possible null/alternative pairs:

$$H_0$$
: $\mu \ge \mu_0$

$$H_{\rm a}: \ \mu < \mu_0$$

$$H_0$$
: $\mu \leq \mu_0$

$$H_{\rm a}: \ \mu > \mu_0$$

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

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



Example:

The NY city provides one of the most comprehensive emergency medical services in the world. Operating in a multiple hospital system with approximately 200 mobile medical units, the service goal is to respond to medical emergencies with a mean time of 12 minutes or less.

The director of medical services wants to formulate a hypothesis test that could use a sample of emergency response times to determine whether or not the service goal of 12 minutes or less is being achieved.



Hypothesis	Conclusion & Action
$H_0: \mu <= 12$	The response time goal is met by Emergency Service team. No action required
H _a : μ > 12	The response time goal is not met by Emergency Service team. Appropriate follow up actions are required.

where,

 μ = mean response time for the population of medical emergency response unit.



Few Important Guidelines:

All statistical conclusions are made in reference to the null hypothesis.

We either **reject** the null hypothesis or **fail to reject** the null hypothesis; we **do not accept** the null hypothesis.

If we **reject** the null hypothesis, then we conclude the data supports the alternative hypothesis.

However, if we fail to reject the null hypothesis, it does not mean we have proven the null hypothesis is "true"



Example 1:

Is there anything we can assume to be true?

Yes. 250 ml in container can be assumed to be true

Which hypothesis pair to be selected?

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

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

$$H_0 = 250 ml$$

$$H_a \neq 250 \text{ ml}$$

If the cans are filled properly, then we **fail to reject** the null hypothesis.

We are not saying that we have proven the null. Just our assumption has held up.





Example 1:

According to the United States Department of Agriculture, in 2009 the average farm size in Texas was 2.3 sq. km. Due to large agriculture business in the past decade, the department wants to know if the current (2019) farm size is larger than it was in 2009.

Establish a null and alternative hypothesis.

Solution:

What is our assumption?

We assume that there has been no change in the farm size since 2009.

This is our null hypothesis.

$$H_0: \mu \le \mu_0$$

 $H_a: \mu > \mu_0$

 $H_0 \le 2.3 \text{ sq. km}$

 $H_a > 2.3 \text{ sq. km}$





Since hypothesis tests are based on sample data, we must allow for the possibility of errors.

Type I error: False Positive

This error occurs when we reject the assumption (null hypothesis) when it should not have been rejected.

Type II error: False Negative

When we fail to reject the assumption (null hypothesis) when it should have been rejected.



Example: NY Medical Emergency Unit

Our Hypothesis was:

 $H_0: \mu <= 12$

Our Alternate Hypothesis was:

 $H_a: \mu > 12$



	Population Condition	
Conclusion	H ₀ True (μ <= 12)	H_0 False (H_a True) ($\mu > 12$)
Accept H ₀	True Positive	False Negative Type II Error
Reject H ₀	False Positive Type I Error	True Negative



Example:







Case:

Let's say we have population mean μ for a population under analysis. The hypothesized mean for the population under analysis is μ_0 . Test if the hypothesized mean comes from the same population under analysis.

$$H_0$$
: $\mu = \mu_0$ H_a : $\mu \neq \mu_0$



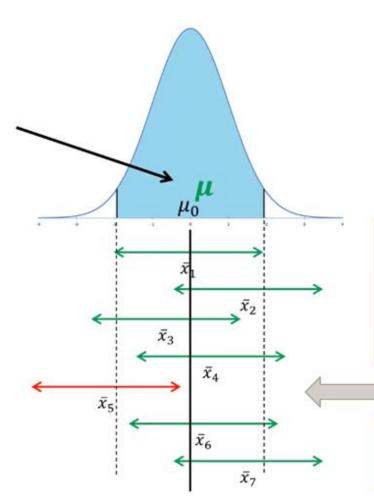
 $\alpha = .05$

95% of all sample means (\bar{x}) are hypothesized to be in this region.

Fail to reject null hypothesis Fail to reject null hypothesis Fail to reject null hypothesis Fail to reject null hypothesis

Reject null hypothesis

Fail to reject null hypothesis Fail to reject null hypothesis



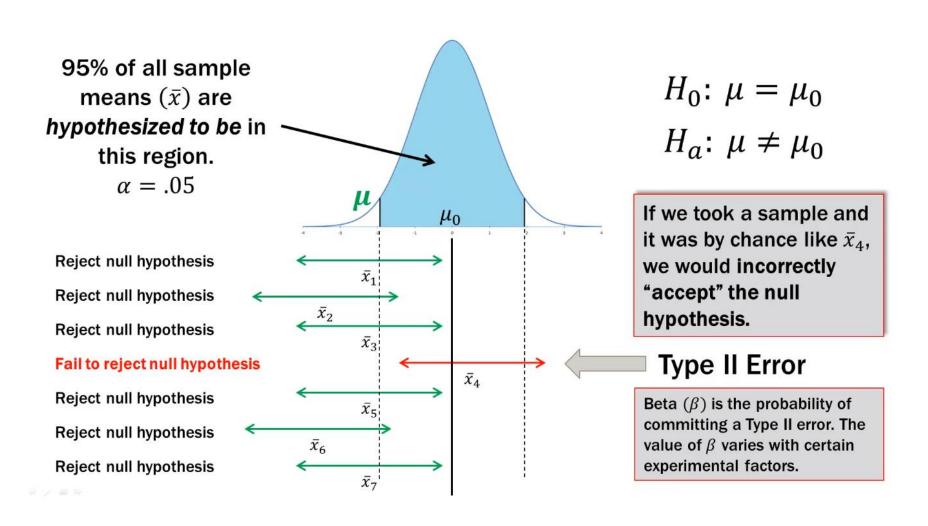
 H_0 : $\mu = \mu_0$ H_a : $\mu \neq \mu_0$

If we took a sample and it was by chance like \bar{x}_5 , we would incorrectly reject the null hypothesis.

Type I Error

 α is the "level of significance" or our tolerance for making a Type I error.



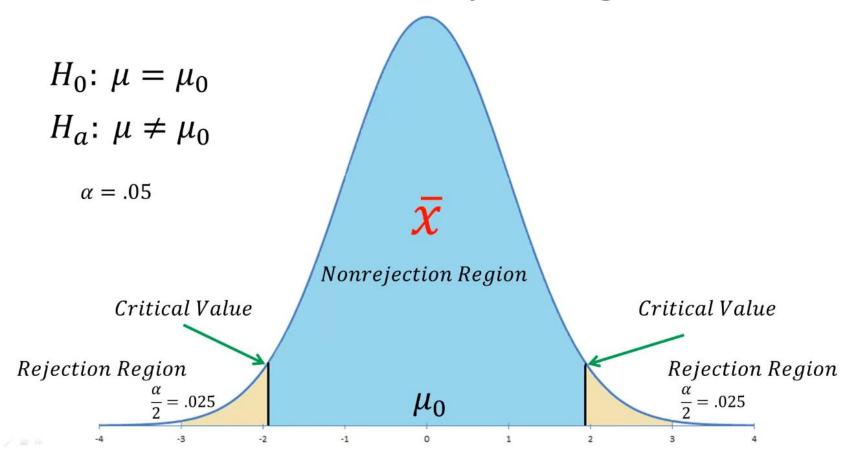


Two Tailed Test Rejection



Two Tailed Test Rejection

The Two-tailed Test Rejection Region



Test of Hypothesis



Test of Hypothesis

Test of Hypothesis

Large Sample Tests

Small Sample Test

Testing
Population
Mean
(One Sample)

Testing
Population
Mean
(Two Sample)

t-test

One Sample
Two Sample

Paired t-

Chisquare test

F - test



Test of Hypothesis

Large Sample Test

- Tests are based on large sample on which hypothesis is test (sample size n at least >30)
- Exact Sampling Distributions are not available.
- Test are based on C.L.T distribution.

Small Sample Test

- Tests are based on small sample on which hypothesis is test
- Exact Sampling Distributions are available.

The **exact sampling distribution** is the sampling distribution of a statistic (estimator, e.g. sample mean x bar) over all the possible sample drawn from a population (finite or infinite).



Large Sample Test



Testing Population Mean (μ): One Sample

When to use?

If you want to test whether or not a given sample is taken from a population whose mean value is specified (μ_0)

How to calculate statistics:

$$z = \frac{\overline{x} - \mu}{\sigma / \sqrt{n}}$$

Where,

 \overline{x} = Sample mean of 'n' observations.

 σ = Population S.D. (known)

 μ_0 = Population mean



Hypothesis on One Sample z test

Here we test whether the given sample belongs to population having mean = μ_0

 μ_0 =Specified value of mean

- H_0 : Mean is equal to μ_0
- H_a : Mean is not equal to μ_0



Testing Population Mean (μ): Two Sample

When to use?

If you want to test whether if two population means are equal or unequal.

How to calculate statistics:

$$z = \frac{(\overline{x_1} - \overline{x_2}) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Where,

 \overline{x} = Sample mean of 'n' observations.

 $\sigma 1$ = Population S.D. (known) of sample one

 μ 1 = Population mean of sample one

2 = Sample mean of 'n2' observations.

 σ 2 = Population S.D. (known) of sample two

 μ 2 = Population mean of sample one



Small Sample Test



Small Sample Test

In small sample test, the sampling distribution that the test statistic follow are majorly classified in three distributions:

- X² distribution (chi-square)
- t-distribution
- F-distribution



Small Sample Test: One Sample t-test

When to use?

If you want to test whether a given sample is coming from a population or not whose mean is specified value and population variance is unknown.

$$t = \frac{\overline{x} - \mu_0}{s / \sqrt{n}}$$

Where,

 \overline{x} = Sample mean of 'n' observations.

s = S.D. of sample of 'n' observations.

 μ_0 = Population mean.



Hypothesis on : One Sample t test

Here we test whether the given sample belongs to population having mean = μ_0

 μ_0 =Specified value of mean

- H_0 : Mean is equal to μ_0
- H_a : Mean is not equal to μ_0



Small Sample Test: Two Sample t-test

When to use?

If you want to test whether given two samples belongs to same population or not. This can be tested by checking equality of means of two samples.

$$t = \frac{\overline{x}_1 - \overline{x}_2}{s_p * \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

Where,

 \overline{x}_1 = Sample mean of n_1 observations

 \overline{x}_2 = Sample mean of n_2 observations

 S_p = Pooled S.D. of two samples



Hypothesis on: Two Sample t test

Here we test whether the given two sample belongs to same population.

- H_0 : Mean of two samples is same.
- H_a: Mean of two samples is not same.
- H_0 : Two samples belongs to same population.
- H_a: Two samples belongs to different population.



Small Sample Test: Paired t-test

When to use?

If you want to check, effectiveness of a new treatment, a new method employed or a training one can apply paired t-test. Here observations on every unit are made before and after applying the treatment or method. Hence if treatment or method is effective, there will be significant difference in observations before applying it and after applying it.

$$t = \frac{\overline{x}_D - \mu_0}{s_D / \sqrt{n}}$$

Where,

 \overline{x}_D = Mean of differences of 'n' paired observations.

 $S_D = S.D.$ of differences of 'n' paired observations.

 μ_0 = Mean of differences of 'n' paired observations under H_0 .



Hypothesis on : Paired t-test

Here we test whether the given two sample belongs to same population or not. This can also be taken as there is no significant difference in samples before and after applying some treatment or method.

- H_0 : Mean of samples before and after treatment is same.
- H_a: Mean of samples before and after treatment is not same.



Small Sample Test: Chi-square test

When to use?

If you have attribute data and if you want to test whether any two attributes are associated with each other or they are independent of each other, we can use Chi-Square test for checking independence of attributes.

$$\chi^2 = \sum \frac{(O_i - E_i)^2}{E_i}$$

Where,

O_i = Observed frequency or Cell count

E_i = Expected frequency or Cell count



Hypothesis on : Chi-Square test

Here we test whether attributes in the given data are independent of each other or not.

- H_0 : Two attributes are independent of each other.
- H_a: Two attributes are not independent of each other.



Small Sample Test: F test

When to use?

If you have two different samples and you want to know whether these two sample have same population variance or not, one can use F test.

$$F = \frac{s_1^2}{s_2^2}$$

Where,

 S_1^2 = Sample variance of first sample of size n_1

 S_2^2 = Sample variance of second sample of size n_2



Hypothesis on : F test

Here we test whether two samples have same variance or not.

- H_0 : Variance ratio is one
- H_a: Variance ratio not equal to one

