

Test of Hypothesis

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Introduction

Test of Hypothesis.

Learned till now: point and interval estimate.

You may encounter a different kind of problem where there are two competing claims about the value of a parameter and you have to determine which one is correct. For example, consider the designing of an air crew escape system. The system consists of an ejection seat and a rocket motor that powers the seat

Example

The rocket motor contains a propellant which should have a mean burning rate of 50 cm/sec in order for the ejection seat to work properly. Too high or low burning rates may lead to possible pilot injury.

The engineering question: Does the mean burning rate of the propellant equal 50 cm/sec, or is it some other value? (higher/lower)?

Statistical Hypothesis

Many types of decision making problems can be converted formulated as hypothesis testing problems.

There is a close connection between hypothesis testing and confidence intervals.

Statistical Hypothesis: A statistical hypothesis is a statement about the parameters of one or more populations.

It may also be thought of as a statement about the probability distⁿ.

Example

Example: Aircor's ~~new~~ escape system.

Consider burning rate as a ~~var~~ r.v.

having a prob. distⁿ. Our interest

is in its mean, a parameter.

Formally,

$H_0: \mu = 50 \text{ cm/sec} \rightarrow \text{null hypothesis}$

$H_1: \mu \neq 50 \text{ cm/sec} \rightarrow \text{alternate "}$

Here it is two-sided alternative Hyp.

$H_0: \mu = 50 \text{ cm/sec}$ | $H_0: \mu = 50 \text{ cm/sec}$

$H_1: \mu < 50 \text{ cm/sec}$ | $H_1: \mu > 50 \text{ cm/sec}$

one-sided alternative Hyp.

Note:

Test of Hypothesis

Test of hypothesis.

A procedure leading to a decision about a ~~parameter~~ particular hypothesis is called a test of hypothesis.

Hyp. testing relies on information gathered from a random sample collected from the population of interest.

Information inconsistent with the H_0 leads to rejection of the hypothesis.

Truth or falsity of H_0 can never be known with certainty.

Testing of Hypothesis: taking a random sample, computing test statistic, make decision

Critical Region

Consider,

$$H_0: \mu = 50 \text{ cm/sec}$$

$$H_1: \mu \neq 50 \text{ cm/sec}.$$

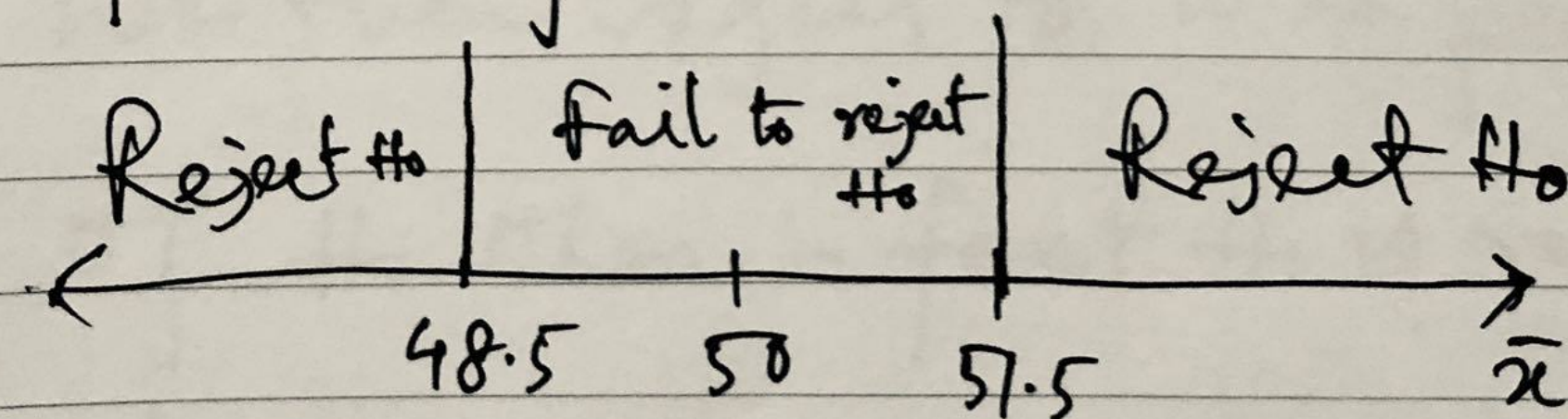
$n = 10$ specimens tested, \bar{x} calculated

Suppose, if $48.5 \leq \bar{x} \leq 51.5$, we will

not reject H_0 . If $\bar{x} < 48.5$ or

$\bar{x} > 51.5$, we will reject H_0 in

favour of H_1 .



Critical Region. | Accepted Region | Critical Region

Errors

This decision procedure may lead to two wrong conclusions.

For example, ~~to~~ suppose true ^{mean} burning rate is really 50 cm/sec, but the randomly selected specimens give an observed value of \bar{x} to be falling into the critical region. We then reject H_0 in ~~to~~ favour of H_1 when in fact H_0 is really true.

Type I error: Rejecting the null Hyp. H_0 when it is true is defined as a Type I error.

Errors

On the other hand, for a different scenario if the true mean mean burning rate is different from 50 cm/sec. yet the sample mean \bar{x} falls in the acceptance region. Then we fail to reject H_0 when it is really false.

Type II error: failing to reject the null hypothesis when it is false is defined as a type II error.

Errors

<u>Decision</u>	<u>H_0 is true</u>	<u>H_0 is false</u>
Fail to reject H_0	No error	Type II error
Reject H_0	Type I error	No error

~~$\alpha = P(\text{Type I error})$~~

$$\alpha = P(\text{Type I error}) = P(\text{reject } H_0 \text{ when } H_0 \text{ is true}).$$

↓
(significance level)

(α -error)
(size of the test)

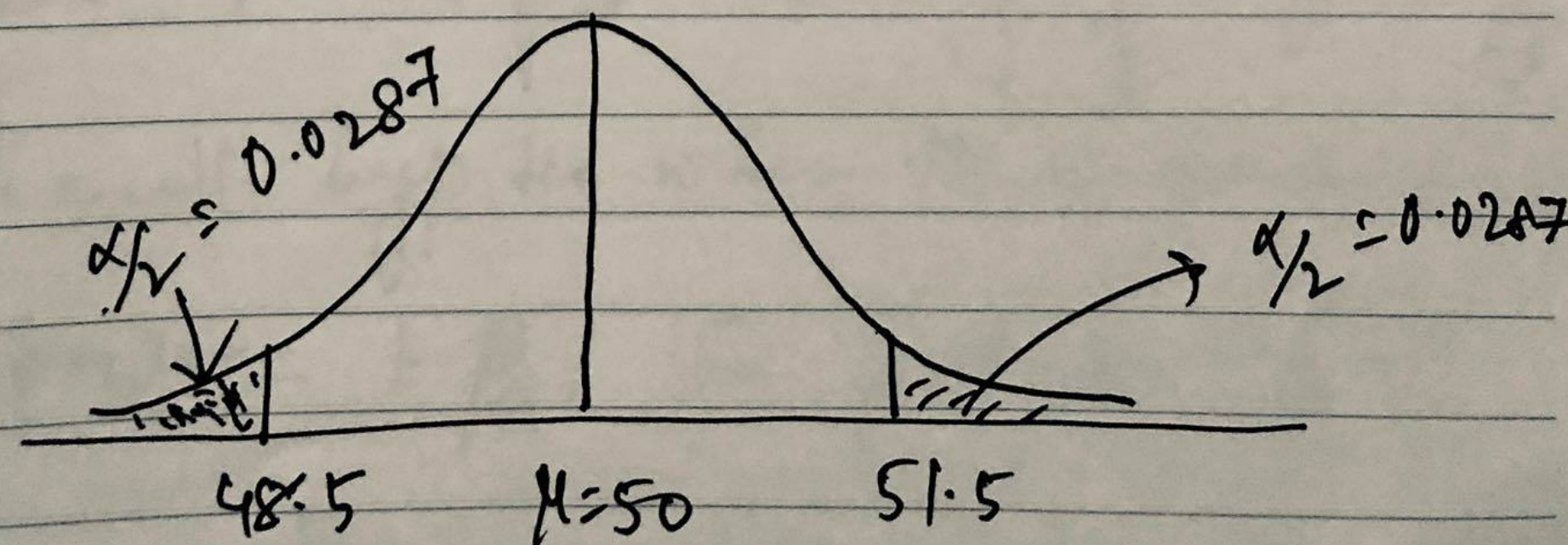
$$\beta = P(\text{Type II error}) = P(\text{fail to reject } H_0 \text{ when } H_0 \text{ is false}).$$

↓
 β -error.

Errors

Note: (A) The size of critical region and consequently the probability of Type I error α , can always be reduced by appropriate selection of the critical values.

(B) Type I error and Type II error are related. A decrease in the probability of one results in the increase in the probability of the other, provided sample size remains same.



Power

Generally, α is controlled when an analyst selects critical values.

Since, the analyst can directly control α , the probability of wrongly rejecting H_0 , rejection of H_0 is considered as a strong conclusion.

$$\alpha = \underline{\underline{0.05}} / 0.1 / 0.01.$$

Power: The power of a statistical test is the probability of rejecting the null hyp. H_0 when H_1 is true.

power = $1 - \beta$, the probability of correctly rejecting a false H_0 .

P-values

P-values

One way to report the results of Hyp. test is to state that the null hypothesis was or was not rejected at a specified α -value or level of significance.

This fixed significance level approach has some disadvantages.

The P-value is the prob. that the test statistic will take on a value that is at least as extreme as the observed ~~as~~ value of the statistic when H_0 is true. P-value conveys much info about the weight of evidence against H_0 , so a decision maker can draw a conclusion at any specified level of significance.

Example

P-value: The P-value is the smallest level of significance that would lead to rejection of the null hypothesis H_0 with the given data.

Let, P-value = 0.038.

Then, H_0 would be rejected at any level of significance greater than or equal to 0.038.

$$\begin{aligned} \text{P-value} &= 1 - P(48.7 < \bar{X} < 51.3) \\ &= 1 - P\left(\frac{48.7 - 50}{2.5/\sqrt{16}} < Z < \frac{51.3 - 50}{2.5/\sqrt{16}}\right) \\ &= 0.038. \end{aligned}$$