

Statistical Hypothesis: A statistical hypothesis is an assumption/assertion/statement about a popⁿ parameter. This assumption may or may not be true. [also about the values of several parameters/about form of an entire prob dist]

Simple Hypothesis: The hypothesis which completely specifies the distⁿ of the parameters is called simple hypothesis.

for example;

$H_0: \theta = \theta_0$ vs $H_1: \theta = \theta_1$ is a simple hypothesis.

Composite Hypothesis: The hypothesis which completely doesn't specify the distⁿ of the parameters is called composite hypothesis.

example;

$H_0: \boxed{\theta = \theta_0}$ vs $H_1: \theta > \theta_0$ or $\theta < \theta_0$ or $\theta \neq \theta_0$

There are two contradictory hypotheses.

(1) Null, (2) Alternative

Ans 4. Null Hypothesis: The null hypothesis is a statement about the popⁿ parameters which is to be tested. It is denoted by H_0 .

Alternative Hypothesis: It is a complementary hypothesis to the null; i.e. the logical opposite of null hypothesis is called alternative hypothesis. It is denoted by H_1 or H_A .

One tailed test: A statistical hypothesis test where the alternative is one sided such as

$$H_0: \mu \leq \mu_0 \text{ vs } H_1: \mu > \mu_0$$

$$\text{or } H_0: \mu \geq \mu_0 \text{ vs } H_1: \mu < \mu_0$$

Two tailed test: A statistical hypothesis test where the alternative hypothesis is two sided is called two tailed test.

Example: $H_0: \theta = \theta_0$ vs $H_1: \theta \neq \theta_0$

$H_1: [\theta \neq \theta_0 \text{ since } < \theta_0 \text{ want } > \theta_0 \text{ or } \text{critically}]$

07/03/17

mean

* A manufacturer claims that the thickness of the spearmint gum it produces is 2.5 cm in a inch. A quality control specialist regularly check this claim. From the production run, he took a R.S. of pieces of gum and measured their thickness. [continuous variable]

$$H_0: \mu = 2.5$$

[μ - parameter usually mean must denote.]

$$H_1: \mu \neq 2.5$$

[many others may normally distributed]

* A biologist was interested in determining whether a new treatment resulted in a lower average height of sunflower than the standard height of 15.2 cm

$$H_0: \mu = 15.2 \quad [\text{Null Hypothesis}]$$

$$\text{vs } H_1: \mu < 15.2 \quad [\text{Alternative Hypothesis}]$$

Null Hypothesis - 'equal' [H₀],
Alt Hypothesis - 'Alternative'

* The scores of a n.s. of 8 students on hypothesis test are as follows -

60, 62, 65, 69, 70, 72, 75 and 78

One researcher wants to test whether the average is greater than 65. \rightarrow Popn. H_1

$$H_0: \mu = 65 \quad [\text{Null Hypothesis}]$$

$$\text{vs } H_1: \mu > 65$$

Reject H_0 | H_0 is true = type 1 error

Accept H_0 | H_0 is false = type 2 error

The error of rejecting null hypothesis when null hypothesis is true, is called type 1 error.

The error of accepting null hypothesis when null hypothesis is false, is called type 2 error.

A usually fixed value (say 5%)

α = level of significance = $P(\text{rejecting } H_0 \mid H_0 \text{ is true})$

The prob[†] of type 1 error is called level of significance.

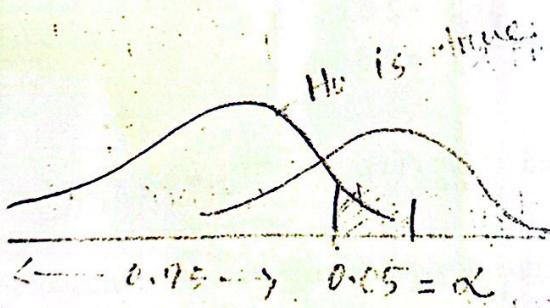
The prob[†] of type 2 error is called ~~level~~

$\beta = P(\text{accepting } H_0 \mid H_0 \text{ is false})$

	H_0 is True	H_0 is False	
Reject H_0	Type I error	Correct α	power of $(1-\beta)$
Don't reject H_0	Correct $1-\alpha$	Type II Error β	

Let, $H_0: \mu = 20$

v/s $H_1: \mu > 20$



Power of the test: The probability of rejecting a false hypothesis is called power of the test.

$$P(\text{reject } H_0 | H_0 \text{ is false}) = 1 - \beta$$

08/03/17

Among Type I error and Type II error - which is more serious ?? Ans; depends on situation suppose.

H_0 : no disease

H_1 : disease present

$$\alpha = \Pr[\text{reject } H_0 \mid H_0 \text{ true}]$$

$\rightarrow +ve \mid \text{no disease}$

Medical Test

$\rightarrow -ve \mid \text{disease} \rightarrow \text{Type I}$

(more serious)

H_0 : innocent

H_1 : criminal

$$\beta = \Pr[\text{accept } H_0 \mid H_1 \text{ true}]$$

$\rightarrow \text{Type II}$

Judge decision

$\rightarrow \text{Punishment} \mid \text{didn't commit murder}$

$\rightarrow \text{no punishment} \mid \text{committed murder}$

Central Limit Th^m:

Let, x_1, x_2, \dots, x_n denote the observations of a r.s. from a distⁿ that has mean μ and variance $\sigma^2 (< \infty)$. Then the r.v.

$$y_n = \frac{\sum_{i=1}^n x_i - n\mu}{\sqrt{n} \cdot \sigma} = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} = \sqrt{n}(\bar{x} - \mu)/\sigma \text{ has}$$

a standard normal distⁿ when n is large.

Test Statistic

Procedures of a test:

- ① state the null and alternative hypothesis
- ② choose the level of significance
- ③ state the assumptions such as normally distributed or equal variance
- ④ Collect data
- ⑤ choose appropriate test statistic based on assumptions.
- ⑥ construct acceptance/rejection region or calculate p-value
- ⑦ Based on ⑤ ^{and} ⑥, take a decision.

Test Statistic: A test statistic is calculated from sample data during a hypothesis test.

It is used to determine whether to reject the null hypothesis. It is also used to

calculate procedure P-Value

Ex To test $H_0: \mu = \mu_0$ vs $H_1: \mu \neq \mu_0$
test statistic, $Z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \sim N(0, 1)$

Mean + Mode = Median

(t -dist) / standard normal

normal distn

symmetric about mean

Normal dist
symmetric about zero

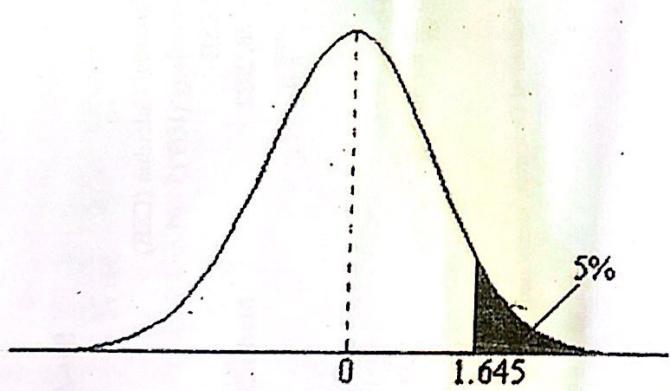
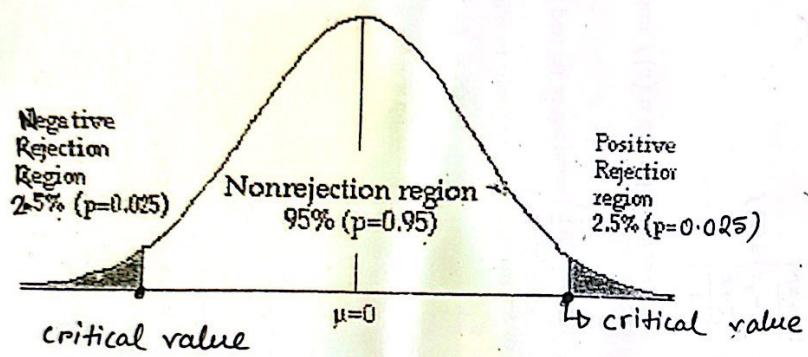
ie failed upper tail (reject $H_0: \mu \geq \mu_0$) 1.645

ie failed(lower tail) (reject $H_0: \mu \leq \mu_0$) -1.645

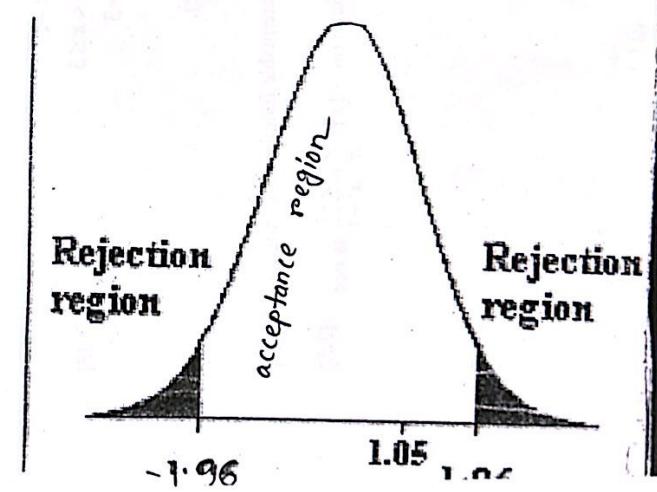
Region of rejection: From the sample data, the researcher computes a test statistic. The specified range of values in which the statistic falls within and leads the researcher to reject the null hypothesis is called region of rejection (critical region).

Region of acceptance: If the statistic falls within a specified range of values that leads the researcher to accept the null hypothesis is called the region of acceptance.

Critical value: The particular value point which separates the rejection region from the acceptance region is called the critical value.



(a) One-tailed test

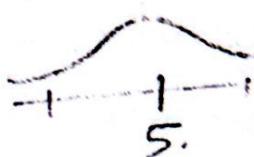


H_0 True হিসেবে শুরু করোন -

i) (যেখানে H_0 Null এবং অন্য একটি অন্য ফল
বিকল আছে,

$$H_0 : \mu = 5$$

$$H_A : \mu > 5$$

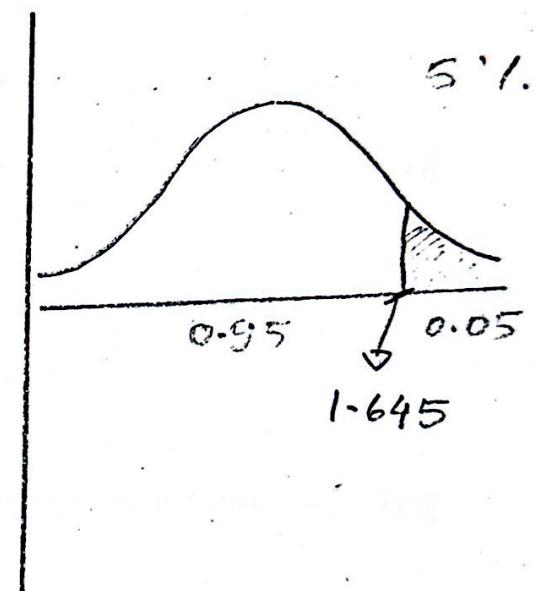


Critical region: From the data, one computes a test statistic. If the value of the test statistic falls within a specified range of values; the null hypothesis (H_0) is rejected; the region constituted by those specified range of values is called critical region.

$$H_0: \mu = \mu_0$$

vs $H_1: \mu > \mu_0$

$$z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$



P-value

1.675

P-value < 0.05, H_0 rejected

P-value > 0.05, H_0 can not be rejected.

P-value = $P[z \geq z_{cal}]$ (one tailed) [when $H_1: \mu > \mu_0$]

P-value = $P[z > z_{cal}] + P[z < -z_{cal}]$ (two tailed)

= $P[z > |z_{cal}|] + P[z < -|z_{cal}|]$

[0.05 is $\frac{1}{2}\alpha = 0.025$ compare with 0.025 because $\alpha = 5\%$.]

P-value : calculated value $\tau_{25(25)}$
extreme value areas $25\% 75\%$,

P-value: The probability of getting the observed or more extreme value when H_0 is true, is called P-value.

$$H_0: \mu = \mu_0$$

(two tailed)

$$H_1: \mu \neq \mu_0$$

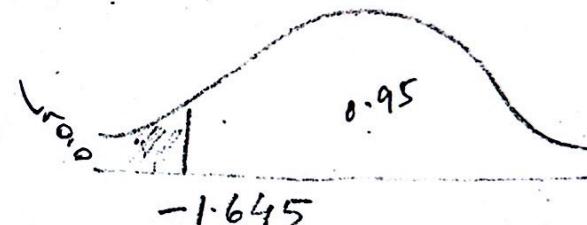
$$\alpha = 5\%$$



$$H_0: \mu = \mu_0$$

$$H_1: \mu < \mu_0$$

$$\alpha = 5\%$$

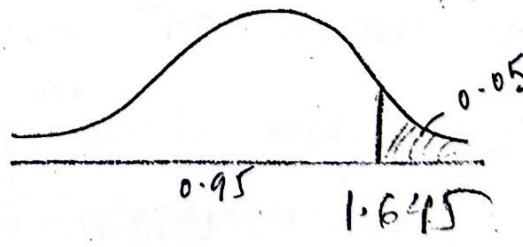


$$P\text{-value} = P[\bar{Z} \leq -|Z_{\text{cal}}|]$$

$$H_0: \mu = \mu_0$$

$$H_1: \mu > \mu_0$$

$$\alpha = 5\%$$



$$P\text{-value} = P[Z \geq |Z_{\text{cal}}|]$$

Tests of Significance

After collecting data (sample), statistical inference requires to make assumptions about the popⁿ from which sample is drawn. This particular assumption depends on the statistical test being used. In view of requirements, significant tests can be used broadly classified to fall into one of the two groups.

- i) Parametric Test,
- ii) Non-parametric Test

If prob^y distⁿ of the popⁿ is required and known, then parametric tests are used. If the prob^y distⁿ is unknown, then non-parametric tests are used. The power of ~~tests~~ non-parametric tests are lower than parametric tests. So, parametric tests should be used if possible.

$$\text{Power} = P[\text{reject } H_0 \mid H_0 \text{ is (+)ve}] ?$$

Uniformly most powerful test (UMP test)

An UMP test is a hypothesis test that has the greatest power among all possible tests of a given size α . [say $\alpha = 0.05$]

* Single Mean Test

im① (i) Popⁿ is normal with known variance.

Suppose x_1, x_2, \dots, x_n be a r.s. of size n which is drawn from a normal pop with mean μ and known variance. Let us define

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

~~Hypothesis:~~ $H_0: \mu = \mu_0$ vs $H_1: \mu \neq \mu_0$

Under the above assumption, the test statistic is given by -

$$Z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$$

standardized test statistic

Decision

accept H_0 if $|Z| < z_{\alpha/2}$

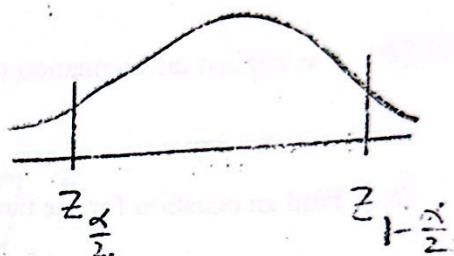
reject H_0

Level of significance = α

Decision:

Reject H_0 if $Z_{\text{cal}} < Z_{\alpha/2}$ or $Z_{\text{cal}} > Z_{1-\alpha/2}$

→ Failed test.



$$\begin{aligned} \text{P-value: } & P[Z > |Z_{\text{cal}}|] + P[Z < -|Z_{\text{cal}}|] \\ &= 2 * P[Z > |Z_{\text{cal}}|] \\ &= 2 * \Theta[1 - P[Z \leq |Z_{\text{cal}}|]] \end{aligned}$$

One-tailed test

case (2) $H_0: \mu = \mu_0$ vs $H_1: \mu > \mu_0$

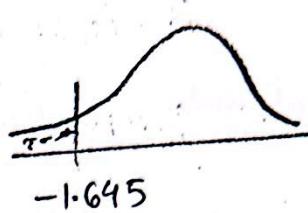
Decision: Reject H_0 if $|Z_{\text{cal}}| > Z_{1-\alpha}$

$$\text{P-value} = P[Z > |Z_{\text{cal}}|] = 1 - P[Z < |Z_{\text{cal}}|]$$



Case (3) $H_0 : \mu = \mu_0$ vs

$H_1 : \mu < \mu_0$



Decision: Reject H_0 if

$$|Z_{\text{cal}}| < Z_\alpha$$

$$\text{P. value} = P[Z \leq -|Z_{\text{cal}}|]$$

2nd situation

~~(1)~~) Popⁿ is normal with variance unknown and sample size > 30

Suppose x_1, x_2, \dots, x_n be n independent random sample from normal popⁿ with mean μ and unknown variance σ^2 .

Let us define $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$ and $s = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$

~~e-1)~~ To test: $H_0 : \mu = \mu_0$ vs $H_1 : \mu \neq \mu_0$

Test statistic $Z = \frac{\bar{x} - \mu_0}{\frac{s}{\sqrt{n}}}$

level of significance $= \alpha$

Decision: Reject H_0 if $|Z_{cal}| > Z_{1-\alpha/2}$ or

$$Z_{cal} < -Z_{\alpha/2}$$

$$P\text{-value} = P[Z > |Z_{cal}|] + P[Z < -|Z_{cal}|]$$

$$= 2 * P[Z \geq |Z_{cal}|]$$

$$= 2 * [1 - P[Z < |Z_{cal}|]]$$

(3rd situation)

Popⁿ is normal with variance unknown and sample size ≤ 30

Suppose: x_1, x_2, \dots, x_n be n i.i.d. r.s. from normal popⁿ with mean μ and unknown variance σ^2 .

Let us define, $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$ and $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$

To test -

$$H_0: \mu = \mu_0 \text{ vs } H_1: \mu \neq \mu_0$$

test statistic $t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \sim t_{n-1}$

level of significance $= \alpha$

Decision:

Reject H_0 if $|t_{cal}| > t_{n-1, 1-\frac{\alpha}{2}}$ or
- upper tail

$$|t_{cal}| < t_{n-1, \frac{\alpha}{2}}$$

$$P\text{-value} = P[t \geq |t_{cal}|] + P[t < -|t_{cal}|]$$

$$= 2 * P[t > |t_{cal}|] = 2 * [1 - P[t \leq |t_{cal}|]]$$

Problem: A new test was designed to have a mean of 80 and SD 10. A r.s. of 20 students take the test and the mean score turns out to be 85. Does this score significantly differ from 80?

Sol: Hence, $\bar{x} = 85$, $\sigma = 10$, $n = 20$

We are to test $H_0: \mu = 80$ vs
 $H_1: \mu \neq 80$

test statistic, $Z = \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} = \frac{85 - 80}{\frac{10}{\sqrt{20}}} = \frac{5}{\sqrt{5}} = 2.236$

$\sim N(0,1)$ H₀ True under H₁ True

Let, level of significance $\alpha = 0.05$

critical value $Z_{\alpha/2} = -1.96$ and $Z_{1-\alpha/2} = 1.96$

Decision: Since, $Z_{\text{cal}} = 2.236 > Z_{1-\frac{\alpha}{2}} = 1.96$
so we may reject H₀ at 5% level of significance.

$$P\text{-value} = 2 * [1 - P[Z \leq Z_{\text{cal}}]] = 2 * 0.0127 = 0.025$$

one

since $p\text{-value} = 0.025 < \alpha = 0.05$ we may
reject H_0 at 5% level of significance.

Hence, this score significantly differs from
80.

Two Means Test (z test, P-value Test)

The common test statistic is

$$\text{Test statistic} = \frac{\bar{x} - \bar{y} - D_0}{\text{standard error}}$$

where \bar{x} and \bar{y} are sample means of two populations and D_0 is the value of difference under H_0 .

Here standard error changes depending on whether the pop'n are -

- (i) independent with known variance
- (ii) " unknown but assumed equal variances
- (iii) independent with unknown and unequal variances.
- (iv) dependent \rightarrow same individual (2 obsⁿ for F-test)

i) independent with known variances: Let

x_1, x_2, \dots, x_n and y_1, y_2, \dots, y_m be two independent random samples from $N(\mu_1, \sigma_1^2)$ and $N(\mu_2, \sigma_2^2)$ respectively. Here σ_1^2 and σ_2^2 are known.

Let us define,

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad \text{and} \quad \bar{y} = \frac{1}{m} \sum_{i=1}^m y_i$$

① Hypothesis: $H_0: \mu_x - \mu_y = D_0$

vs. $H_1: \mu_x - \mu_y \neq D_0$

Test statistic, $z = \frac{\bar{x} - \bar{y} - D_0}{\text{S.E. } (\bar{x} - \bar{y})}$

$$= \frac{\bar{x} - \bar{y} - D_0}{\sqrt{\frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}}}$$

$$\begin{aligned} \text{S.E. } (\bar{x} - \bar{y}) &= \sqrt{V(\bar{x}) + V(\bar{y})} \\ &= \sqrt{2 \text{Cov}(\bar{x}, \bar{y})} \\ &= \sqrt{\frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}} \end{aligned}$$

Let, α = Level of significance

Decision: Reject H_0 if $z_{\text{cal}} \leq z_{\alpha/2}$ or

$$z_{\text{cal}} \geq z_{1-\alpha/2}$$

$$\begin{aligned}
 P\text{-value} &:= 2 * P[Z \geq |Z_{\text{calc}}|] \\
 &= 2 * [1 - P[Z \leq |Z_{\text{calc}}|]] \\
 &= 2 * [1 - P[Z \leq |Z_{\text{calc}}|]]
 \end{aligned}$$

Case ② $H_0: \mu_x - \mu_y \leq D_0$
 $H_1: \mu_x - \mu_y > D_0$

Decision: Reject H_0 if $Z_{\text{calc}} > Z_{1-\alpha}$

$$P\text{-value} = P[Z > |Z_{\text{calc}}|] = 1 - P[Z \leq |Z_{\text{calc}}|]$$

Case ③ $H_0: \mu_x - \mu_y \geq D_0$
 $H_1: \mu_x - \mu_y < D_0$

Decision: Reject H_0 if $Z_{\text{calc}} < Z_{\alpha}$

$$P\text{-value} = P[Z \leq |Z_{\text{calc}}|]$$

Ex. One reason for wage differentials bet' men and women is the difference of their work experience. Given data of experiences as-

Dec

<u>Men</u>	<u>Women</u>
$n_x = 100$	$n_y = 80$
$\bar{x} = 13.5$	$\bar{y} = 11.5$
$\sigma_x = 5.2$	$\sigma_y = 3.8$

Test the hypothesis that, "men tend to have er at least 4.5 more years of experiences than women".

$D_0 = 0$
2.50

\leq

$$H_0: \mu_x - \mu_y \leq 4.5$$

$$\text{vs. } H_1: \mu_x - \mu_y > 4.5$$

$$Z = \frac{\bar{x} - \bar{y} - D_0}{\text{SE}(\bar{x} - \bar{y})} = \frac{13.5 - 11.5 - 4.5}{\sqrt{\frac{5.2^2}{100} + \frac{3.8^2}{80}}} = -3.72$$

.0005

<0.01

Let, $\alpha = 0.05$. Then $Z_{\alpha} = -1.645$
 $= Z_{0.05}$

since $Z_{\text{cal}} = -3.72 < Z_{0.05} = -1.645$;

so we may reject H_0 at 5% level of significance. (Ans) $P(Z \leq -3.72) = \text{P-value}$

$$\begin{aligned} \text{P-value} &= P[Z < -|Z_{\text{cal}}|] / 1 - P[Z \leq -3.723] \\ &= P[Z < -3.72] = 1 - P[Z \leq 3.723] \\ &= \text{pnorm}(-3.72) = 0.0000957 \\ &= <0.001 \end{aligned}$$

(2) Independent with unknown but assumed equal variances.

Let, x_1, x_2, \dots, x_n and y_1, y_2, \dots, y_m be two samples from $N(\mu_1, \sigma_1^2)$ and $N(\mu_2, \sigma_2^2)$ where σ_1^2 and σ_2^2 are unknown but $\sigma_1^2 = \sigma_2^2$

Let us define -

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \text{ and } \bar{y} = \frac{1}{m} \sum_{i=1}^m y_i$$