

Case 1 Testing Hypothesis of Normal distribution when σ is known (given)

\therefore The test we will opt is Z-test

The concerned test statistic is $Z = \frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma} \sim N(0,1)$
under $H_0: (\mu = \mu_0)$

Given significance level is α .

Depending on H_1

$\hookrightarrow H_1: \mu > \mu_0$, then we reject H_0 if

$$Z_{obs} > Z_{\alpha}$$

$\hookrightarrow H_1: \mu < \mu_0$, then we reject H_0 if $Z_{obs} < -Z_{\alpha}$

$\hookrightarrow H_1: \mu \neq \mu_0$, then we reject H_0 if

$$Z_{obs} < Z_{1-\alpha/2} = Z_{\alpha/2} \quad [\because Z \text{ is symm}]$$

$$\text{or } Z_{obs} > Z_{\alpha/2}$$

$$\text{ie } |Z_{obs}| > Z_{\alpha/2}$$

Case 2 Testing for mean for Normal distribution when σ is unknown

Let X follow a distribution $N(\mu, \sigma^2)$ where σ^2 is not known.

Let X_1, X_2, \dots, X_n be n iid random samples drawn from the distribution of X .

$$\text{Let } s^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \text{ where } \bar{X} = \frac{1}{n} \left(\sum_{i=1}^n X_i \right)$$

Let $H_0: \mu = \mu_0$ & given significance level be α .

Consider the test-statistic $t = \frac{\sqrt{n}(\bar{X} - \mu_0)}{s}$ under H_0 .

$$\text{Now we see } \left(\frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} \right) \sim Z(0,1)$$

$$\text{and } \frac{s^2}{\sigma^2} = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \sim \chi_{n-1}^2$$

$$\therefore t = \frac{\frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma}}{\frac{s}{\sigma}} \sim \frac{Z(0,1)}{\sqrt{\chi_{n-1}^2}} = t_{n-1}$$

$$\Rightarrow t = \frac{\sqrt{n}(\bar{X} - \mu_0)}{s} \sim t_{n-1}$$

Now depending on H_1

$\rightarrow H_1: \mu < \mu_0$, we reject H_0 if

$$t_{obs} < t_{n-1, \alpha}$$

$\rightarrow H_1: \mu > \mu_0$, we reject H_0 if

$$t_{obs} > t_{n-1, \alpha}$$

$\rightarrow H_1: \mu \neq \mu_0$, we reject H_0 if

$$|t_{obs}| > t_{n-1, \alpha/2}$$

Case 3: Testing for variance of a normal population whose mean is known

Let X follow normal distribution with mean μ_0
 $X \sim N(\mu_0, \sigma^2)$ where σ^2 is being tested

Consider n iid random samples X_1, X_2, \dots, X_n drawn from the distribution of X .

Let $H_0: \sigma = \sigma_0$

Consider $S^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \mu_0)^2$

Then we define

Then under H_0 , consider test-statistic,

$$t = \frac{nS^2}{\sigma_0^2} \sim \chi_n^2$$

Depending on H_1 , at α -level of significance

\rightarrow If $H_1: \sigma > \sigma_0$, we reject H_0 if

$$t_{obs} > \chi_{n, \alpha}^2$$

\rightarrow If $H_1: \sigma < \sigma_0$, we reject H_0 if

$$t_{obs} < \chi_{n, 1-\alpha}^2$$

\rightarrow If $H_1: \sigma \neq \sigma_0$, we reject H_0 if

$$t_{obs} < \chi_{n, 1-\alpha/2}^2$$

$$t_{obs} > \chi_{n, \alpha/2}^2$$

$$\frac{(n-1)S^2}{\sigma_0^2}$$

case 1: Testing for variance of a normal population when mean is unknown.

let X follow a normal distribution $N(\mu, \sigma^2)$ where μ is unknown & σ^2 is being tested for.

let n iid random variables, be X_1, X_2, \dots, X_n drawn from the distribution of X .

\therefore let $H_0: \sigma = \sigma_0$

Now consider the test statistic $t = \frac{(n-1)S^2}{\sigma_0^2}$ under H_0

$$\text{where } S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \text{ \& }$$

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

under H_0 , $t \sim \chi_{n-1}^2$

Thus depending on H_1 , at α l.o.s,

\hookrightarrow If $H_1: \sigma < \sigma_0$, we will reject H_0 if

$$t_{obs} < \chi_{n-1; 1-\alpha}^2$$

\hookrightarrow If $H_1: \sigma > \sigma_0$, we will reject H_0 if

$$t_{obs} > \chi_{n-1; \alpha}^2$$

\hookrightarrow If $H_1: \sigma \neq \sigma_0$, we will reject H_0 if

$$t_{obs} > \chi_{n-1; \alpha/2}^2$$

or

$$t_{obs} < \chi_{n-1; 1-\alpha/2}^2$$