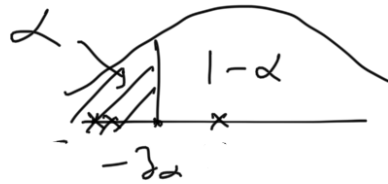


Testing of statistical hypothesis :

(1) r.v. $X \sim N(\mu, \sigma^2)$ i.s. X_1, \dots, X_n Sample mean \bar{X}
 $\swarrow \quad \nwarrow$
 unknown \quad known ($\sigma^2 = \sigma_0^2$)

H_0	H_1	C (Critical region) Or rejection region for H_0
$\mu \geq \mu_0$	$\mu < \mu_0$	$Z_0 = \frac{\bar{X} - \mu_0}{\sigma_0/\sqrt{n}} < -z_\alpha$
$\mu \leq \mu_0$	$\mu > \mu_0$	$Z_0 = \frac{\bar{X} - \mu_0}{\sigma_0/\sqrt{n}} > z_\alpha$
$\mu = \mu_0$	$\mu \neq \mu_0$	$ Z_0 > z_{\frac{\alpha}{2}}$



(2) $X \sim N(\mu, \sigma^2)$ i.s. X_1, \dots, X_n $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$ $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$
 \swarrow
 unknown

H_0	H_1	C
$\mu \geq \mu_0$	$\mu < \mu_0$	$t_0 = \frac{\bar{X} - \mu_0}{S/\sqrt{n}} < -t_{\alpha, n-1}$
$\mu \leq \mu_0$	$\mu > \mu_0$	$t_0 > t_{\alpha, n-1}$
$\mu = \mu_0$	$\mu \neq \mu_0$	$ t_0 > t_{\frac{\alpha}{2}, n-1}$

(3) $X \sim N(\mu, \sigma^2)$
 \swarrow
 unknown

H_0

H_1

C

$$\begin{array}{lll}
 \sigma^2 \leq \sigma_0^2 & \sigma^2 > \sigma_0^2 & \chi_0^2 = \frac{(n-1)S^2}{\sigma_0^2} > \chi_{\alpha, n-1}^2 \\
 \sigma^2 \geq \sigma_0^2 & \sigma^2 < \sigma_0^2 & \chi_0^2 < \chi_{1-\alpha, n-1}^2 \\
 \sigma^2 = \sigma_0^2 & \sigma^2 \neq \sigma_0^2 & \chi_0^2 > \chi_{\frac{\alpha}{2}, n-1}^2 \text{ or } \chi_0^2 < \chi_{1-\frac{\alpha}{2}, n-1}^2
 \end{array}$$

—X—

Two samples

(i) $X_1 \sim N(\mu_1, \sigma_1^2)$ or X_{11}, \dots, X_{1n_1} \bar{X}_1 S_1^2
indep $\left\{ \begin{array}{l} \text{unknown} \\ \text{known} \end{array} \right.$
 $X_2 \sim N(\mu_2, \sigma_2^2)$ or X_{21}, \dots, X_{2n_2} \bar{X}_2 S_2^2

$H_0: \mu_1 = \mu_2$ vs $H_1: \mu_1 \neq \mu_2$

C $|Z_0| = \left| \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \right| > z_{\frac{\alpha}{2}}$

(ii) $X_1 \sim N(\mu_1, \sigma_1^2)$ or X_{11}, \dots, X_{1n_1} \bar{X}_1 S_1^2
indep $\left\{ \begin{array}{l} \text{unknown} \end{array} \right.$
 $X_2 \sim N(\mu_2, \sigma_2^2)$ or X_{21}, \dots, X_{2n_2} \bar{X}_2 S_2^2

(ii)_T $\sigma_1^2 = \sigma_2^2 = \sigma^2$ $H_0: \mu_1 = \mu_2$ vs $H_1: \mu_1 \neq \mu_2$

C $|t_0| > t_{\frac{\alpha}{2}, n_1+n_2-2}$

$t_0 = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{S^2}{n_1} + \frac{S^2}{n_2}}} \sim t_{n_1+n_2-2}$

$$s_p = \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

$$S_p^2 = \frac{(n_1-1)S_1^2 + (n_2-1)S_2^2}{n_1 + n_2 - 2}$$

(ii) $\sigma_1^2 \neq \sigma_2^2$ $H_0: \mu_1 = \mu_2$ vs $H_1: \mu_1 \neq \mu_2$

C $|t_0| > t_{\frac{\alpha}{2}, v}$

$$t_0 = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}} \sim t_v$$

$$v = \frac{\left(\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}\right)^2}{\frac{(S_1^2/n_1)^2}{n_1+1} + \frac{(S_2^2/n_2)^2}{n_2+1}} - 2$$

Testing equality of variance

$H_0: \sigma_1^2 = \sigma_2^2$ vs $H_1: \sigma_1^2 \neq \sigma_2^2$

Under H_0

$$F_0 = \frac{\frac{(n_1-1)S_1^2}{(n_1-1)\cancel{\sigma_1^2}}}{\frac{(n_2-1)S_2^2}{(n_2-1)\cancel{\sigma_2^2}}} = \frac{S_1^2}{S_2^2} \sim F_{n_1-1, n_2-1}$$

C $F_0 > F_{\frac{\alpha}{2}, n_1-1, n_2-1}$ or $F_0 < F_{1-\frac{\alpha}{2}, n_1-1, n_2-1}$

D

paired t-test
 $(X_{11}, X_{21}), \dots, (X_{1n}, X_{2n})$ ^{indep}

$$H_0: \mu_1 = \mu_2 \text{ vs } H_1: \mu_1 \neq \mu_2 \quad \frac{D_i = X_{1i} - X_{2i}}{i=1, \dots, n}$$

$$\equiv H_0: \mu_D = 0 \text{ vs } H_1: \mu_D \neq 0 \quad \bar{D}, S_D^2$$

$$\mu_D = \mu_1 - \mu_2$$

✓
 $C \quad |t_0| > t_{\frac{\alpha}{2}, n-1}$

$$t_0 = \frac{\bar{D}}{S_D / \sqrt{n}}$$

—x—

$$p\text{-value} \leq P(Z > Z_0)$$

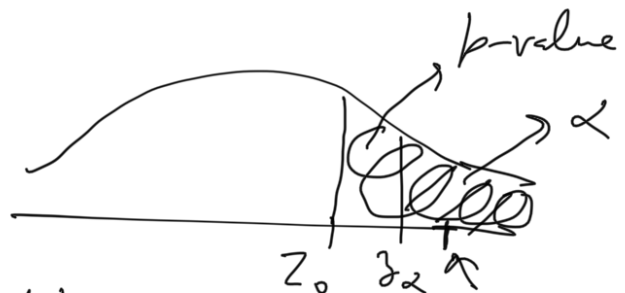
$$H_0: \mu \leq \mu_0 \text{ vs } H_1: \mu > \mu_0$$

$C \quad \underline{Z_0 > z_\alpha}$

$$p\text{-value} < \alpha \rightarrow \text{reject } H_0$$

$$p\text{-value} > \alpha \rightarrow \text{accept } H_0$$

—x—



Contingency table test

Row	Columns			
	1	2	...	C
1	O_{11}	O_{12}	...	O_{1C}
2	O_{21}	O_{22}	...	O_{2C}

H_0 : rows & columns members
of classification
are indep.

$$\begin{matrix} & & \dots & & \\ & & \dots & & \\ n & O_{n1} & O_{n2} & \dots & O_{nc} \end{matrix}$$

$$p_{ij} = u_i v_j$$

$$\text{mle} \quad \hat{u}_i = \frac{1}{n} \sum_{j=1}^c O_{ij}, \quad \hat{v}_j = \frac{1}{n} \sum_{i=1}^n O_{ij}$$

assuming independence, the expected # in each cell is

$$E_{ij} = n \hat{u}_i \hat{v}_j = \frac{1}{n} \sum_{m=1}^c O_{im} \sum_{k=1}^n O_{kj}$$

for layer n

$$\chi^2_o = \sum_{i=1}^n \sum_{j=1}^c \frac{(O_{ij} - E_{ij})^2}{E_{ij}} \sim \chi^2_{(n-1)(c-1)}$$

$$C \quad \chi^2_o > \chi^2_{\alpha, (n-1)(c-1)}$$