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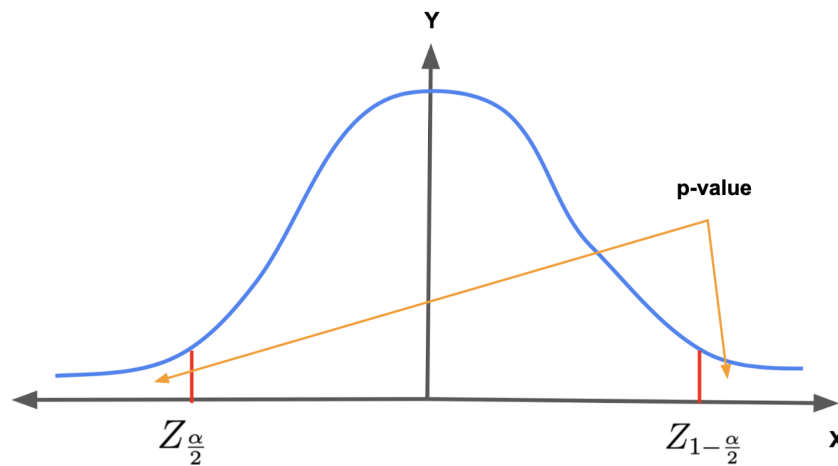
Hypothesis Test

Consider the observations $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} D_1(\mu_1, \sigma_1)$, $Y_1, \dots, Y_n \stackrel{\text{iid}}{\sim} D_2(\mu_2, \sigma_2)$.

With X_i independent of $Y_j \quad \forall (i \in \{1, \dots, n\}, j \in \{1, \dots, m\})$.

i. Equality:

$$H_0 : \mu_1 = \mu_2 \Leftrightarrow \mu_1 - \mu_2 = 0 \quad \text{vs} \quad H_a : \mu_1 \neq \mu_2 \Leftrightarrow \mu_1 - \mu_2 \neq 0$$



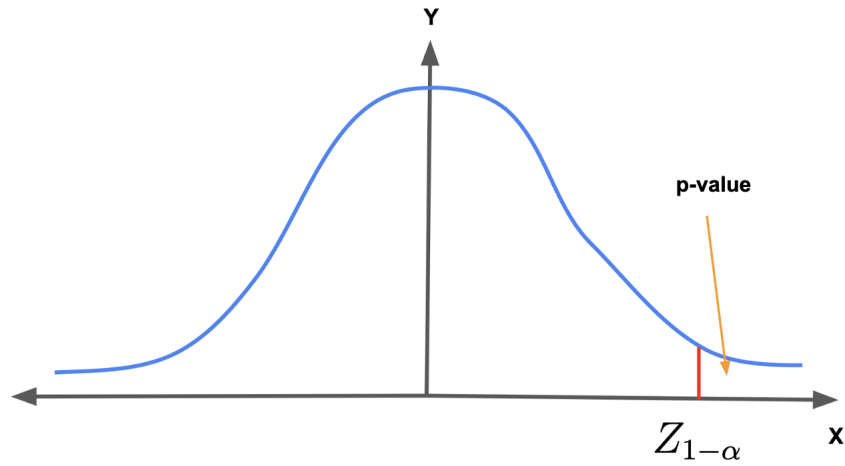
$$T = \frac{(\bar{x} - \bar{y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{S_x^2}{n} + \frac{S_y^2}{m}}}$$
$$T_{|H_0} = \frac{(\bar{x} - \bar{y})}{\sqrt{\frac{S_x^2}{n} + \frac{S_y^2}{m}}}$$

If $T_{|H_0} > Z_{1-\frac{\alpha}{2}}$ (**critical value**) or $T_{|H_0} < Z_{\frac{\alpha}{2}}$ (**critical value**) then we **can reject** null hypothesis.

Else we can't reject null hypothesis.

ii. Greater:

$$H_0 : \mu_1 > \mu_2 \Leftrightarrow \mu_1 - \mu_2 > 0 \quad \text{vs} \quad H_a : \mu_1 \leq \mu_2 \Leftrightarrow \mu_1 - \mu_2 \leq 0$$



$$T = \frac{(\bar{x} - \bar{y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{S_x^2}{n} + \frac{S_y^2}{m}}}$$

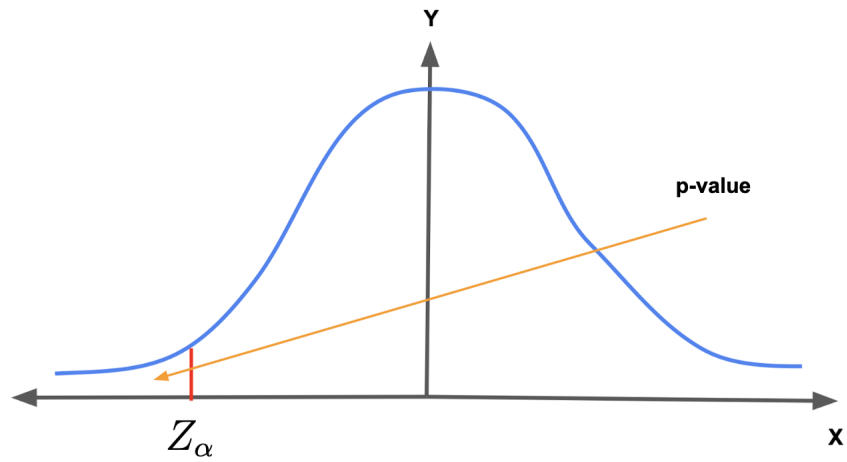
$$T_{|H_0} = \frac{(\bar{x} - \bar{y})}{\sqrt{\frac{S_x^2}{n} + \frac{S_y^2}{m}}}$$

If $T_{|H_0} \leq Z_{1-\alpha}$ (**critical value**) then we **can reject** null hypothesis.

Else we **can't reject** null hypothesis.

iii. Lower:

$$H_0 : \mu_1 < \mu_2 \Leftrightarrow \mu_1 - \mu_2 < 0 \quad \text{vs} \quad H_a : \mu_1 \geq \mu_2 \Leftrightarrow \mu_1 - \mu_2 \geq 0$$



$$T = \frac{(\bar{x} - \bar{y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{S_x^2}{n} + \frac{S_y^2}{m}}}$$

$$T_{|H_0} = \frac{(\bar{x} - \bar{y})}{\sqrt{\frac{S_x^2}{n} + \frac{S_y^2}{m}}}$$

If $T_{|H_0} \geq Z_\alpha$ (**critical value**) then we **can reject** null hypothesis.

Else we **can't reject** null hypothesis.

Demonstrations

$$T_1 = \frac{1}{n} \cdot \sum_{i=1}^n X_i$$

$$T_2 = \frac{1}{m} \cdot \sum_{j=1}^m Y_j$$

$$Z = T_1 - T_2$$

$$\mathbb{E}(Z) = \mathbb{E}(T_1 - T_2) = \mathbb{E}(T_1) - \mathbb{E}(T_2)$$

$$\begin{aligned} &= \mathbb{E}\left(\frac{1}{n} \cdot \sum_{i=1}^n X_i\right) - \mathbb{E}\left(\frac{1}{m} \cdot \sum_{j=1}^m Y_j\right) = \frac{1}{n} \cdot \sum_{i=1}^n \mathbb{E}(X_i) - \frac{1}{m} \cdot \sum_{j=1}^m \mathbb{E}(Y_j) \\ &= \frac{1}{n} \cdot \sum_{i=1}^n \mu_1 - \frac{1}{m} \cdot \sum_{j=1}^m \mu_2 = \frac{1}{n} \cdot [n \cdot \mu_1] - \frac{1}{m} \cdot [m \cdot \mu_2] \\ &= \mu_1 - \mu_2 \end{aligned}$$

\Rightarrow

$$\mathbb{E}(Z) = \mu_1 - \mu_2$$

$$\begin{aligned} \mathbb{V}(Z) &= \mathbb{V}(T_1 - T_2) \stackrel{\text{ind}}{=} (1)^2 \cdot \mathbb{V}(T_1) + (-1)^2 \cdot \mathbb{V}(T_2) = \mathbb{V}(T_1) + \mathbb{V}(T_2) \\ &= \mathbb{V}\left(\frac{1}{n} \cdot \sum_{i=1}^n X_i\right) + \mathbb{V}\left(\frac{1}{m} \cdot \sum_{j=1}^m Y_j\right) \stackrel{\text{ind}}{=} \frac{1}{n^2} \cdot \sum_{i=1}^n \mathbb{V}(X_i) + \frac{1}{m^2} \cdot \sum_{j=1}^m \mathbb{V}(Y_j) \\ &= \frac{1}{n^2} \cdot \sum_{i=1}^n (\sigma_1)^2 + \frac{1}{m^2} \cdot \sum_{j=1}^m (\sigma_2)^2 = \frac{1}{n^2} \cdot [n \cdot (\sigma_1)^2] + \frac{1}{m^2} \cdot [m \cdot (\sigma_2)^2] \\ &= \frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m} \end{aligned}$$

\Rightarrow

$$\mathbb{V}(Z) = \frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}$$