

5) Given $D_1 = \{x_1, x_2, \dots, x_n\}$ be iid from $\text{Normal}(\mu_1, \sigma_1^2)$

$D_2 = \{y_1, y_2, \dots, y_m\}$ be iid from $\text{Normal}(\mu_2, \sigma_2^2)$

x 's and y 's are independent and $\mu_1, \sigma_1^2, \mu_2, \sigma_2^2$ are unknown

then the hypothesis is

$$H_0: \mu_1 > \mu_2 \quad H_1: \mu_1 \leq \mu_2$$

using unpaired T-test with threshold value $\delta > 0$ to check hypothesis

$$T = \frac{\bar{D}}{\sqrt{\frac{S_x^2}{n} + \frac{S_y^2}{m}}} \quad \text{where } D = \frac{\sum_{i=1}^n x_i}{n} - \frac{\sum_{i=1}^m y_i}{m} = \bar{x} - \bar{y}$$

Since this is a one-sided test, to validate null hypothesis we need

$$T > -\delta$$

Assume n & m are large

using CLT we get $\bar{x} \sim \text{Nor}\left(\mu_1, \frac{\sigma_1^2}{n}\right)$ and

$$\bar{y} \sim \text{Nor}\left(\mu_2, \frac{\sigma_2^2}{m}\right)$$

$$\therefore \bar{D} = \bar{x} - \bar{y} \sim \text{Nor}\left(\mu_1 - \mu_2, \frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}\right)$$

As σ_1^2 and σ_2^2 are unknown, we can replace them with their plugin estimators.

$$\bar{D} \sim \text{Nor} \left(\mu_1 - \mu_2, \frac{s_x^2}{n} + \frac{s_y^2}{m} \right)$$

$$\begin{aligned} \Pr(\text{Type I error}) &= \Pr(\text{reject } H_0 \mid H_0 \text{ true}) \\ &= P(T < -\delta) \end{aligned}$$

$$= \Pr \left(\frac{\bar{D}}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} < -\delta \right)$$

$$= \Pr \left(\bar{D} < -\delta \sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}} \right)$$

$$= \Pr \left(\bar{D} - (\mu_1 - \mu_2) < -\delta \sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}} - (\mu_1 - \mu_2) \right)$$

$$= \Pr \left(\frac{\bar{D} - (\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} < -\delta - \frac{(\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} \right)$$

$$= \Phi \left(-\delta - \frac{(\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} \right)$$

$$Pr(\text{Type I error}) = \Phi \left(-\delta - \frac{(\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} \right)$$

$$Pr(\text{Type II error}) = Pr(\text{Accept } H_0 \mid H_1 \text{ true})$$

$$= Pr(T > -\delta)$$

$$= Pr \left(\frac{\bar{D}}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} > -\delta \right)$$

Applying same steps as of type I error

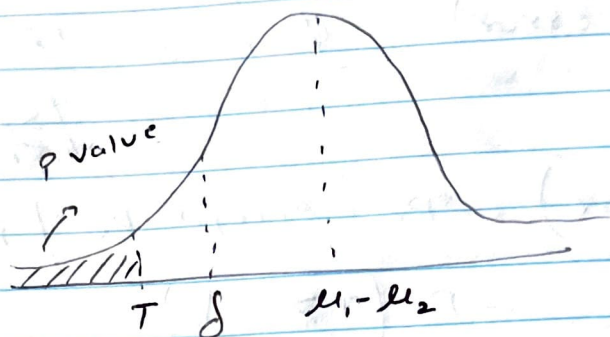
$$Pr(T > -\delta) = Pr \left(\frac{\bar{D} - (\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} > -\delta - \frac{(\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} \right)$$

$$= 1 - \Phi \left(-\delta - \frac{(\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}} \right)$$

5b) For the unpaired test we get T-Statistic as

$$T = \frac{\bar{X} - \bar{Y}}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{m}}}$$

If we reject H_0 , then $T < -\delta$



$$P\text{-value} = \Pr \left(\frac{\bar{D}}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{n}}} < \frac{\bar{x} - \bar{y}}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{n}}} \right)$$

$$= \Pr(\bar{D} < \bar{x} - \bar{y})$$

$$= \Pr \left(\frac{\bar{D} - \mu_1 - \mu_2}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{n}}} < \frac{(\bar{x} - \bar{y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{n}}} \right)$$

$$= \Pr \left(Z < \frac{(\bar{x} - \bar{y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{n}}} \right)$$

$$P\text{-value} = \Phi \left(\frac{(\bar{x} - \bar{y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_x^2}{n} + \frac{s_y^2}{n}}} \right)$$