

Comparing Two Independent Samples

The sampling distribution for ...

$$\bar{X}_1 \text{ is approximately Normal with Mean} = \mu_1, \quad SD = \frac{\sigma_1}{\sqrt{n_1}}$$

$$\bar{X}_2 \text{ is approximately Normal with Mean} = \mu_2, \quad SD = \frac{\sigma_2}{\sqrt{n_2}}$$

$$\bar{X}_1 - \bar{X}_2 \text{ is approx Normal with Mean} = \mu_1 - \mu_2, \quad SD = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

$$z_s = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{SD_{\bar{X}_1 - \bar{X}_2}} \rightarrow t_s = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{SE_{\bar{X}_1 - \bar{X}_2}}$$

Example: 30 mice given one of two diets for 21 days.

Reduction in cholesterol in mg/dl is measured on day 22

Diet	N	Mean	StDev
Bean	15	26.46	5.90
Oat	15	32.23	9.56

$$H_o: \mu_1 - \mu_2 = D_0$$

$$t_s = \frac{(\bar{X}_1 - \bar{X}_2) - D_0}{SE_{\bar{X}_1 - \bar{X}_2}}$$

$$\sigma_1 = \sigma_2 \rightarrow t_s = \frac{(\bar{X}_1 - \bar{X}_2) - D_0}{\sqrt{s_p^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \text{ with } df = n_1 + n_2 - 2 \quad (\text{Equal Variance } T\text{-test})$$

Rejection Region at the α level of significance:

$$\begin{aligned} \text{Reject } H_o \text{ in favor of } & H_a: \mu_1 - \mu_2 > D_0 \text{ if } t_s \geq t_{\alpha} \\ & H_a: \mu_1 - \mu_2 < D_0 \text{ if } t_s \leq -t_{\alpha} \\ & H_a: \mu_1 - \mu_2 \neq D_0 \text{ if } |t_s| \geq t_{\alpha/2} \end{aligned}$$

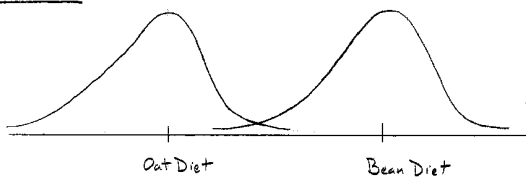
$$\sigma_1 \neq \sigma_2 \rightarrow t_s = \frac{(\bar{X}_1 - \bar{X}_2) - D_0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \quad (\text{Unpooled or Unequal Variance } T\text{-test})$$

Satterthwaite's approximation for the degrees of freedom:

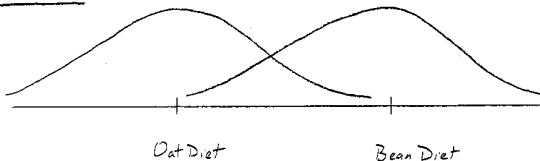
$$df = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2} \right)^2}{\frac{\left(\frac{s_1^2}{n_1} \right)^2}{(n_1 - 1)} + \frac{\left(\frac{s_2^2}{n_2} \right)^2}{(n_2 - 1)}}$$

Individual Sampling Distributions for $\bar{X}$

Situation 1



Situation 2



Non-parametric Tests for 2 Independent Samples

- Test $H_o: \tilde{\mu}_1 = \tilde{\mu}_2$ as opposed to $H_o: \mu_1 = \mu_2$
- Use recoded data: the ranks of the observations as opposed to the original values.
- Do not rely on the assumption of normally distributed populations $\rightarrow$ Nonparametric

A T-test on the Ranks

- Rank the combined data from both samples (ties get assigned the average rank).
- Do a 2-sample t -test on the ranks, instead of the observed values.
 - This test is roughly equivalent to the Wilcoxon Rank Sum Test.

Wilcoxon Rank Sum Test (a.k.a. Mann-Whitney Test)

- Rank the combined data from both samples (ties get assigned the average rank).
- T = Sum of the ranks in sample 1 (called W or S in some software).

When H_o is true, the sampling distribution for T is approx. normal (if $n_1 > 10$ & $n_2 > 10$) with

- Mean: $\mu_T = \frac{n_1(n_1 + n_2 + 1)}{2}$
- Variance: $\sigma_T^2 = \frac{n_1 n_2}{12} (n_1 + n_2 + 1)$
- Test Statistic: $z_s = \frac{T - \mu_T}{\sigma_T}$

$$\begin{aligned} \text{Rejection Region at the } \alpha \text{ level of significance: } & H_a: \tilde{\mu}_1 > \tilde{\mu}_2 \text{ if } z_s \geq z_{\alpha} \\ & H_a: \tilde{\mu}_1 < \tilde{\mu}_2 \text{ if } z_s \leq -z_{\alpha} \\ & H_a: \tilde{\mu}_1 \neq \tilde{\mu}_2 \text{ if } |z_s| \geq z_{\alpha/2} \end{aligned}$$

When there are ties in the ranks there is a correction to the variance:

$$\sigma_T^2 = \frac{n_1 n_2}{12} (n_1 + n_2 + 1) - \left(\frac{n_1 n_2}{12} \frac{\sum_{j=1}^k t_j(t_j^2 - 1)}{(n_1 + n_2)(n_1 + n_2 - 1)} \right)$$

JMP:

- (Analyze: Fit Y by X: <red> Nonparametric: Wilcoxon Test)
- When there are more than 2 samples, this is known as the Kruskal-Wallis Test