

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 → Nonparametric

(version 1) 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.

(version 2) Wilcoxon Rank Sum Test (a.k.a. Mann-Whitney Test)

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

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

- Mean: $\mu_w = \frac{n_1(n_1 + n_2 + 1)}{2}$
- Variance: $\sigma_w^2 = \frac{n_1 n_2}{12} (n_1 + n_2 + 1)$
- Test Statistic: $z_s = \frac{W - \mu_w}{\sigma_w}$

W should be S in the 3 formulas to the left

Rejection Region at the α level of significance:

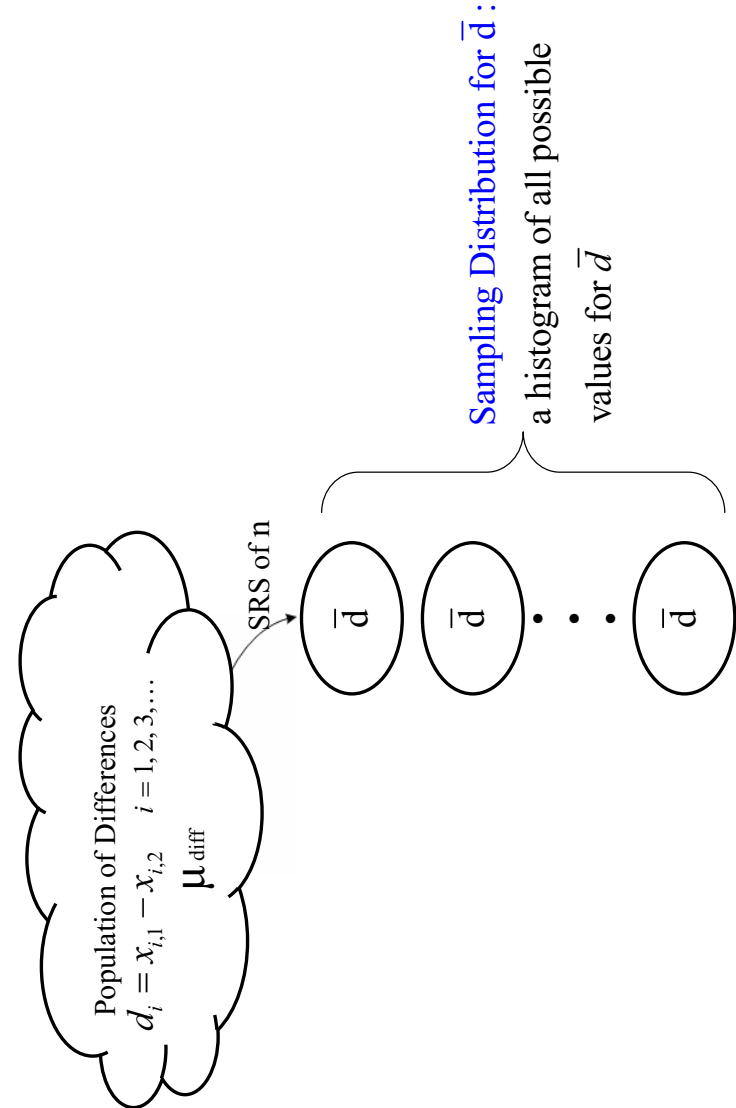
$H_a : \tilde{\mu}_1 > \tilde{\mu}_2$ if $z_s \geq z_\alpha$
 $H_a : \tilde{\mu}_1 < \tilde{\mu}_2$ if $z_s \leq -z_\alpha$
 $H_a : \tilde{\mu}_1 \neq \tilde{\mu}_2$ if $|z_s| \geq z_{\alpha/2}$

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

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

R:

- plac = c(0.9, 0.37, 1.63, 0.83, 0.95, 0.78, 0.86, 0.61, 0.38, 1.97)
- alco = c(1.46, 1.45, 1.76, 1.44, 1.11, 3.07, 0.98, 1.27, 2.56, 1.32)
- `wilcox.test(plac, alco, paired=FALSE, correct=F, exact=F)`
- When there are more than 2 samples, this is known as the Kruskal-Wallis Test, `kruskal.test()` in R



T-Test for 2 Paired Samples

- Each individual in sample 1 is matched to one individual in sample 2 in order to control for other factors and reduce variability

$$\circ d_i = x_{i,1} - x_{i,2} \quad i = 1, 2, 3, \dots, n_d$$

- $H_o : \mu_d = 0 \rightarrow t_s = \frac{\bar{d} - 0}{s_d / \sqrt{n_d}}$ with $df = n_d - 1$

- $100(1 - \alpha)\%$ Confidence Interval for μ_d

$$\bar{d} \pm t_{\frac{\alpha}{2}, n_d - 1} \left(\frac{s_d}{\sqrt{n_d}} \right)$$

Example:

Research question: Does oral contraceptive (OC) use increase SBP?

X = Systolic Blood Pressure (SBP)

$$d_i = x_{i,(no-OC)} - x_{i,(yes-OC)}$$

$$H_o : \mu_d = 0$$

SBP_no_OC	SBP_yes_OC	diff
115	128	-13
112	115	-3
107	106	1
119	128	-9
115	122	-7
138	145	-7
126	132	-6
109	109	0
104	102	2
115	117	-2

Non-parametric Tests for a single Sample

- Do not rely on the assumption of normally distributed populations → Nonparametric

Wilcoxon Signed Rank Test

- Compute the observe differences, $d_i = x_{i1} - x_{i2}$, for the n observed pairs.
- Remove all values with $d_i = 0$
- let n^* be the number of non-zero values.
- Rank the *absolute values* of the differences (ties get assigned the average rank).
 V_+ = Sum of the positive ranks
 V_- = Sum of the negative ranks
- $V = \text{minimum}(V_+, V_-)$

When H_o is true, the sampling distribution for V is approx. normal (if $n > 30$) with

- Mean: $\mu_V = \frac{n^*(n^* + 1)}{4}$
- Variance: $\sigma_V^2 = \frac{n^*(n^* + 1)(2n^* + 1)}{24}$
- Test Statistic: $z_s = \frac{V - \mu_V}{\sigma_V}$

Rejection Region at the α level of significance:

$$H_a : \tilde{\mu}_d > 0 \quad \text{if} \quad z_s \geq z_\alpha$$

$$H_a : \tilde{\mu}_d < 0 \quad \text{if} \quad z_s \leq -z_\alpha$$

$$H_a : \tilde{\mu}_d \neq 0 \quad \text{if} \quad |z_s| \geq z_{\alpha/2}$$

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

$$\sigma_V^2 = \frac{n^*(n^* + 1)(2n^* + 1)}{24} - \frac{1}{2} \sum_{j=1}^k t_j(t_j - 1)(t_j + 1)$$

R:

- sbp1 = c(115, 112, 107, 119, 115, 138, 126, 109, 104, 115)
sbp2 = c(128, 115, 106, 128, 122, 145, 132, 109, 102, 117)
diff = sbp1 - sbp2
- wilcox.test(diff, paired=TRUE, correct=F, exact=F) #OR#
wilcox.test(sbp1, sbp2, paired=TRUE, correct=F, exact=F)

Approximate Sample Size Calculations for 2-Sample Tests (assuming $n_1 = n_2$)

1) For a desired Margin of Error (E):

a) 2 Independent Samples: $(\bar{x}_1 - \bar{x}_2) \pm t_{\alpha/2} \sqrt{\frac{\hat{\sigma}_1^2}{n_1} + \frac{\hat{\sigma}_2^2}{n_2}}$

$$E \approx z_{\alpha/2} \sqrt{\frac{\hat{\sigma}_1^2}{n} + \frac{\hat{\sigma}_2^2}{n}} \rightarrow n \approx \frac{(z_{\alpha/2})^2 (\hat{\sigma}_1^2 + \hat{\sigma}_2^2)}{E^2}$$

b) Paired Samples: $\bar{d} \pm t_{\alpha/2} \frac{\hat{\sigma}_d}{\sqrt{n_d}}$

$$E \approx z_{\alpha/2} \frac{\hat{\sigma}_d}{\sqrt{n_d}} \rightarrow n_d \approx \frac{(z_{\alpha/2})^2 \hat{\sigma}_d^2}{E^2}$$

2) For testing H_0 with a fixed Type I and Type II error probability (α and β)

a) 2 Independent Samples: vs. $H_a: \mu_1 - \mu_2 = \Delta$

$$n \approx \frac{(\hat{\sigma}_1^2 + \hat{\sigma}_2^2) (z_\alpha + z_\beta)^2}{\Delta^2} \quad (z_{\alpha/2} \text{ for a 2-sided test})$$

b) Paired Samples: vs. $H_a: \mu_d = \Delta$

$$n \approx \frac{\hat{\sigma}_d^2 (z_\alpha + z_\beta)^2}{\Delta^2} \quad (z_{\alpha/2} \text{ for a 2-sided test})$$

Ex: (Mouse diets)

Find the number of observations that would be needed to have $power = 90\%$ for rejecting H_0 in favor of $H_a: \mu_{Bean} - \mu_{Oat} = 7$ at the $\alpha=0.05$ level (1-sided test).

Descriptive Statistics: Bean, Oat

Variable	N	Mean	StDev	Variance
Bean	15	26.46	5.90	34.81
Oat	15	32.23	9.56	91.45

In a prospective study individuals are followed for a period of time to see if a “condition” develops.

- a.k.a. Longitudinal study or follow-up study

In a retrospective study individuals with & without a “condition” are asked about past exposures (or current status).

- a.k.a. Cross-sectional study

Example:

A study of the relationship between oral contraceptive use (OC) and blood pressure (BP) in premenopausal women.

- Longitudinal study
 - Identify a group of nonpregnant, women of childbearing age & not current OC users.
 - Measure BP (*baseline measurement*).
 - Re-screen the women after 1 year to determine a subgroup that have become OC users. This subgroup is the *study group*.
 - Measure BP at the follow-up visit
 - Compare the *baseline* and *follow-up* BP of the women in the study group.
- Cross-sectional study
 - Identify a group of OC users and a group of non-OC users among nonpregnant, women of childbearing age.
 - Measure the BP of each woman.
 - Compare the BP of the OC and non-OC users.