


p. 1


Recall: Question  $\rightarrow$  Hypotheses  $\rightarrow$  Data Collection  $\rightarrow$  Data Analysis  $\rightarrow$  Inference about the larger population

For one population, z test for  $\sigma$  known  
t  $\sigma$  unknown samples

I. If we have two populations: + independent  $\mu_1, \mu_2$



$\mu_1$   
 $\sigma_1$



$\mu_2$   
 $\sigma_2$

$\Rightarrow$ 

$\bar{x}_1$ 
 $\bar{x}_2$ 

{
}

two random samples

We are interested in the difference in means  $(\mu_1 - \mu_2)$  for which the best estimator is  $(\bar{x}_1 - \bar{x}_2)$

$H_0: (\mu_1 - \mu_2) = 0$

For one pop., focus was only  $\sigma$  known vs unknown. Here need to understand if  $\sigma_1^2 = \sigma_2^2$  or not.

Two approaches:

1. Test assuming  $=$  + then  $\neq$  + compare.
2. Rule of thumb: If one of the s values is four or more times the other, assume unequal variance.

$\rightarrow$  If  $s_1 \geq 4s_2$   $\rightarrow$  unequal  
and  $\sigma_1^2 = \sigma_2^2$

For two pop., we need to compute a pooled variance = the weighted average of the two sample variances:  $s_p^2$

$$s_p^2 = \frac{[(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2]}{(n_1 + n_2 - 2)}$$

We then find our test statistic and df  
 $df = (n_1 + n_2 - 2)$

$$t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{[s_p^2 (1/n_1 + 1/n_2)]}$$

p.2

value from  $H_0$

If  $\sigma_1^2 \neq \sigma_2^2$ ,

$$df_{\neq} = \frac{[(s_1^2/n_1) + (s_2^2/n_2)]^2}{\left[ \frac{(s_1^2/n_1)^2}{(n_1-1)} + \frac{(s_2^2/n_2)^2}{(n_2-1)} \right]}$$

and  $t_{\neq} = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\left( \frac{s_1^2}{n_1} + \frac{s_2^2}{n_2} \right)}}$

larger  $df$ , larger  $n$  have lower prob. of Type II error.

$df = > df_{\neq} \rightarrow$  if in doubt, choose = variance test

\* Remember, z- & t-tests assume a normal distribution!  
 $\rightarrow$  plot the data first!

II. If we have two samples that are somehow linked (before/after, old part/new part, in the wild/in captivity) with two measures on each subject:

	A	B	Diff
1	•	→ •	x
2	•	→ •	x
3	•	→ •	x
4	•	→ •	x
5	•	→ •	x

our statistic of interest is not two two separate measures, but the diff. btwn them  $x_b$   
 $x_d = x_A - x_B$

Our <sup>null</sup> hypothesis then takes the form: p3  
 $H_0: \mu_D = 0$  &  $H_a: \mu_D > < \neq 0$   
 $\mu_D$   
 $\uparrow$  "Matched pairs"  
 "Paired sample"

The test statistic is

$$t_D = \frac{\bar{x}_D - \mu_D}{s_D / \sqrt{n_D}} \quad \text{with } df = n_D - 1$$

(looks like a one population test!)

III. If your data are very non-normal, you can't use tests involving t statistics

→ need a non-parametric test

Wilcoxon Signed Rank Sum Test

for two populations

nonnormal data

matched pairs data

Process: Estimates the differences  $x_D$

Take absolute value of  $x_D$

Eliminate all  $x_D = 0$

Rank the remainder from 1 = smallest to  $n$  = highest value, but separately for + diff. & - diff.

Sum the ranks to get totals

$T_+$  &  $T_-$

If  $T_+$  &  $T_-$  differ significantly, we can infer  $H_0$  of no difference is false

Ex:

$x_1$	$x_2$	$b$	$ D $	Rank +	Rank -
5	6	-1	1		1
7	7	0	0		
9	4	5	5	4	
11	14	-3	3		3
13	15	-2	2		2
				<hr/>	<hr/>
				4	6
				$T_+$	$T_-$

$$T_+ > T_-$$

for samples of  $n > 30$  the totals are approximately normally distributed. with

$$E(T) = \frac{n(n+1)}{4} \quad \sigma_T = \sqrt{\frac{n(n+1)(2n+1)}{24}}$$

$$\text{and } z = \frac{T_+ - E(T)}{\sigma_T}$$

↑  $T_+$  is the test statistic in this case.