

## Friedman Test

• A nonparametric test used to compare several treatments (populations) when all treatments are applied (in random order) to each subject in the sample. (If there is no blocking, use the Kruskal-Wallis test if a nonparametric procedure is called for.)

**Set Up.** Each of  $k$  treatments is independently applied to  $b$  blocks (or subjects) and each outcome  $X_{ij}$  (where  $i = 1, \dots, b$  and  $j = 1, \dots, k$ ) is recorded. (We assume that the results within one block do not influence the results within the other blocks.)

	Treatments			
Blocks	1	2	...	$k$
1	$X_{11}$	$X_{12}$	...	$X_{1k}$
2	$X_{21}$	$X_{22}$	...	$X_{2k}$
⋮	⋮		⋮	
$b$	$X_{b1}$	$X_{b2}$	...	$X_{bk}$

### Hypotheses.

$H_0$ : treatments are similarly effective ( $p_{i1} = p_{i2} = \dots = p_{ic}$ , see below)

$H_1$ : treatments differ in effectiveness

### Obtaining Test Statistic

First, for each block replace  $X_{ij}$  with its rank  $R(X_{ij})$  within that block. Use average ranks in case of ties. Let  $R_j$  denote the sum of ranks for treatment  $j$ .

	Treatments			
Blocks	1	2	...	$k$
1	$R(X_{11})$	$R(X_{12})$	...	$R(X_{1k})$
2	$R(X_{21})$	$R(X_{22})$	...	$R(X_{2k})$
⋮	⋮		⋮	
$b$	$R(X_{b1})$	$R(X_{b2})$	...	$R(X_{bk})$
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Treatment rank sums:	$R_1$	$R_2$	...	$R_k$

Let  $A = \sum_{i=1}^b \sum_{j=1}^k [R(X_{ij})]^2$  and  $C = \frac{bk(k+1)^2}{4}$ . Then Friedman's **test statistic** (adjusted for possible ties) is given by

$$T = \frac{(k-1) \sum_{j=1}^k \left( R_j - \frac{b(k+1)}{2} \right)^2}{A - C},$$

which has an approximate  $\chi^2_{(k-1)}$  distribution under the null hypothesis.

If there are no ties,  $T$  simplifies to  $T = \frac{12}{bk(k+1)} \sum_{j=1}^k \left( R_j - \frac{b(k+1)}{2} \right)^2$ .

Another route that gives a better approximation of the true distribution uses

$$F^* = \frac{(b-1)T}{b(k-1)-T},$$

which has an approximate  $F_{[k-1, (b-1)(k-1)]}$  distribution under the null hypothesis.

The test statistic  $F^*$  can be obtained by two-way ANOVA on the ranks  $R(X_{ij})$ .

**Decision Rule.** Reject  $H_0$  if the test statistic is too large or if the  $p$ -value is too small.

**Multiple Comparisons.** If the null hypothesis is rejected, it is appropriate to ascertain which pairs of treatments differ. Treatments  $i$  and  $j$  are considered different if

$$|R_i - R_j| > t_{1-\alpha/2} \sqrt{\frac{2(bA - \sum R_j^2)}{(b-1)(k-1)}}.$$

**Example.**(from Biomed Central Ltd). Pain scores of five patients receiving four separate treatments.

	Treatment				
Patient (block)	A	B	C	D	
1	6	9	10	16	
2	9	16	16	32	
3	14	14	22	67	
4	10	14	40	19	
5	11	16	17	60	

Table of ranks

	Treatment				
Patient (block)	A	B	C	D	
1	1	2	3	4	
2	1	2.5	2.5	4	
3	1.5	1.5	3	4	
4	1	2	4	3	
5	1	2	3	4	
Sum( $R_j$ )	5.5	10	15.5	19	

$H_0$ : All four treatments are equally effective in treating pain.

$H_1$ : Not all treatments are equally effective in treating pain.

$$A = \sum_{i=1}^b \sum_{j=1}^k [R(X_{ij})]^2 = 1^2 + 2^2 + \dots + 4^2 = 149$$

and  $C = \frac{bk(k+1)^2}{4} = \frac{5(4)(5^2)}{4} = 125$ , thus

$$T = \frac{(k-1) \sum_{j=1}^k \left( R_j - \frac{b(k+1)}{2} \right)^2}{A-C} = \frac{3(7^2 + 2.5^2 + 3^3 + 6.5^2)}{149-125} = 13.3125.$$

with  $p$ -value = .004 [from TI-83:  $\chi^2$ cdf(13.3125, 10^99, 3)].

Or use  $F^* = \frac{(b-1)T}{b(k-1)-T} = \frac{4(13.3125)}{5(3)-13.3125} = 31.56$  with  $p$ -value = .0000056.

Since the null hypothesis is rejected, for multiple comparisons, with  $\alpha = .05$ , we have

$$t_{1-\alpha/2} \sqrt{\frac{2(bA - \sum R_j^2)}{(b-1)(k-1)}} = 2.1788 \sqrt{\frac{2(5*149 - 5.5^2 - 10^2 - 15.5^2 - 19^2)}{3(4)}} = 3.2682.$$

Hence if  $|R_i - R_j| > 3.2682$ , we can say that treatments  $i$  and  $j$  differ.

Here, we see that all the treatments have a statistically significant difference.

**Reference.** *Practical Nonparametric Statistics*, 3rd ed, W.J. Conover, John Wiley & Sons, 1999