

$$MSE = \frac{SSE}{(b-1)(k-1)} = \frac{0.9933277}{6} = 0.1655546$$

$$F_T = \frac{MSTR}{MSE} = \frac{21.523624}{0.1655546} \quad F_B = \frac{MSB}{MSE} = \frac{41.188179}{0.1655546}$$

ANOVA Table

Source of

Variation	df	SS	MS	F
Treatments	2	7.12667	3.563335	21.523624
Blocks	3	21.946669	7.315563	44.188179
Error	6	.9933277	.1655546	
Tot.	11	30.06667		

$$H_0: \mu_1 = \mu_2 = \mu_3$$

$$H_a: \text{at least one } \mu \neq ; \alpha = 0.01$$

$$\text{test stat: } F_T = \frac{MSTR}{MSE} = \frac{21.523624}{0.1655546}$$

$$\text{R.R. - Reject } H_0 \text{ if } F_T > F_{\alpha}(k-1, (b-k)(k-1)) = F_{0.01}(2, 6) = 10.92$$

Since $F_T = 21.52 > 10.92$, we reject H_0 and conclude that @ 1% level of significance there is an evidence that the 3 couriers differ in their speed of delivery.

→ must do a follow-up analysis to find out which couriers differ.

Tukey's h.s.d.: Multiple Comparisons

① Calculate $\binom{k}{2} = \binom{3}{2} = 3$ pairs of $|\bar{y}_i - \bar{y}_j|$
 for $H_0: \mu_i = \mu_j$ $i, j = 1, 2, 3$
 $H_a: \mu_i \neq \mu_j$ $(i \neq j)$

$$\bar{y}_1 = \frac{T_1}{4} = 4.65 \quad \bar{y}_3 = \frac{T_3}{4} = 6.5 \quad \bar{y}_2 = \frac{T_2}{4} = 5.25$$

$$|\bar{y}_1 - \bar{y}_2| = 0.6 < 1.2877879 \Rightarrow \mu_1 = \mu_2 \quad \text{i.e. there are differences in speed of delivery b/w countries A + C}$$

$$|\bar{y}_1 - \bar{y}_3| = 1.85 > 1.2877879 \Rightarrow \mu_1 \neq \mu_3$$

$$|\bar{y}_2 - \bar{y}_3| = 1.25 > 1.2877879 \Rightarrow \mu_2 \neq \mu_3$$

$$\textcircled{2} \text{ h.s.d.} = q_{\alpha} [K, \text{error}] \sqrt{\frac{\text{MSE} \left(\frac{1}{n_i} + \frac{1}{n_j} \right)}{2}} = q_{\alpha} [K, (b-1)(k-1)] \sqrt{\frac{\text{MSE}}{b}}$$

$$= 9.01 [3, 6] \sqrt{\frac{1.65544}{4}} = 6.33 \dots = \underline{1.2877879}$$

If an ass.
or test
encl prob
here

Is blocking needed?

② $H_0: \beta_1 = \beta_2 = \beta_3 = \beta_4$ (i.e. blocking is not needed); $\alpha = 0.1$
 $H_a: \text{at least one } \beta \neq$ (i.e. blocking is beneficial)

only have
to do if
question
asks for
block test:

$$\text{Test-Stat: } F_B \text{ test} = \frac{\text{MSSB}}{\text{MSE}} = \underline{44.188179}$$

$$\text{R.R. We reject } H_0 \text{ if } F_B > F_{\alpha}(b-1, (b-1)(k-1)) = F_{0.1}(3, 6) = 9.78$$

Since $F_B = 44.188179 > 9.78$ we reject H_0 and conclude that @ 1% level of significance there is an evidence that blocking is beneficial, hence R.B.D & can be used

✓ Comparing RBD to CRD

Don't have to do unless asked for in the question

(i.e. Relative Efficiency (RE))
 Let MSE_{RBD} and MSE_{CRD} denote MSE's under RBD/CRD respectively. One way of measuring the precision is to calc. the variance of (\bar{y}_i)

- for RBD the variance of treat. means
 $var(\bar{y}_i) = MSE_{RBD}$

- for CRD $var(\bar{y}_i) = \frac{MSE_{CRD}}{r}$, where $r = \#$ of replications of observ.'s of each treat. that are required to satisfy the following relationship

$$MSE_{RBD} = \frac{MSE_{CRD}}{r}$$

$$\Rightarrow \left(\frac{r}{b}\right) = \frac{MSE_{CRD}}{MSE_{RBD}}$$

variance
Efficiency

each treat. that are required to satisfy the following relationship

$kb = n$

computational form.

$$RE = \frac{b \cdot MSB + b(k-1) MSE_{RBD}}{(kb-1) MSE_{RBD}}$$

Ex

cont'd

$$RE = \frac{3(7.3155563) + 4(2)(.1655546)}{(11)(.165546)} = 12.778594 = 13$$

i.e. approx. 13 times as many observ.'s in each treat. would be needed if we were to use CRD instead of RBD in order to obtain same precision for comparison of treat. means

Chap: 15.7 - Non-Parametric Approach to RBD
(Friedman Rank Test)

- If assumption of normality is not satisfied
- We work w/ ranks
- We rank the observs from smallest to largest w/in each block
- Tied observs will rec. the average rank

Assumptions:

- ① RBD - ~~1 block~~
- ② Each treat.-block combo. has distribution w/ approx the same shape + the same spread
- ③ Blocks + treat. do not interact

$H_0: \mu_{d1} = \mu_{d2} = \dots = \mu_{dk} ; \alpha$
 $H_a: \text{@ least one } \mu_{di} \neq$

Test stat: $F_R = \frac{12}{bk(k+1)} \left[\sum_{i=1}^k T_{ri}^2 \right] - 3b(k+1)$

where: T_{ri} = sum of ranks in each treat.

* Check: $\sum_{i=1}^k T_{ri} = \frac{b}{2} k(k+1)$ # of blocks must be ≥ 5

RB - Reject H_0 if $F_R > \chi_{\alpha}^2; (k-1)$

Ex:

can't

Assume ① RBD ✓

- ② Distributions of each courier + time of day hour approx the same shape + the same spread.
- ③ Couriers + times of day do not interact

Couriers Treatment

Time of Day	A	B	C	Check $\sum T_{ki} = 4+8+12 = 24$
9:30am	3.6 (1)	4.2 (2)	5 (3)	$= \frac{bk(k-1)}{2} = \frac{12(4)}{2} = 24 \checkmark$
11:30am	5.4 (1)	5.8 (2)	7 (3)	
12:30 pm	6.1 (1)	7 (2)	9.1 (3)	
2:00 pm	3.5 (1)	4 (2)	4.9 (3)	
	$T_{k1} = 4$	$T_{k2} = 8$	$T_{k3} = 12$	

from
error

H_0 : No difference in speed (ie. times of delivery for 3 couriers ($m_A = m_B = m_C$))
 H_a : not all of them the same; $\alpha = .01$

Test stat: $F_R = \frac{12}{(4.3)(4)} \left[\frac{\sum_{i=1}^3 T_{ki}^2}{bk(k-1)} \right] - 3b(k-1) =$
 $= \frac{12}{(4.3)(4)} [4^2 + 8^2 + 12^2] - 3(4)(4) =$
 $= .25(224) - 48 = 56 - 48 = 8$

* Check: $\sum_{i=1}^3 T_{ki} = 4 + 8 + 12 = 24$

$$\frac{bk(k-1)}{2} = \frac{12(4)}{2} = 24$$

We reject H_0 if $F_R > X^2_{(k-1); \alpha} = X^2_{(2); .01} = 9.210$

Since our $F_R = 8 < 9.210$, we do not reject H_0 and conclude that @ 1% level of significance there is no evidence to indicate that the delivery times of 3 different couriers are different.

If we used $\alpha = .05 \Rightarrow X^2_{(2); .05} = 5.991$

$\Rightarrow F_p = 8 > 5.991 \Rightarrow$ Reject H_0 and @ 5% level of significance. There are differences among 3 couriers.

- Which couriers are different?

Nemenyi's Procedure

- a post-hoc multiple comparison procedure used after Friedman's test to declare which sample medians are not equal.

- have to make $\binom{k}{2} = \frac{k(k-1)}{2}$ comparisons

① compute average ranks \bar{R}_j for each of the k groups i.e. $\bar{R}_j = \frac{\sum R_i}{n_j}$ ($j=1 \dots k$)

② critical range = $q_{\alpha}(k, \infty) \sqrt{\frac{k(k+1)}{12b}}$

upper-tailed critical value from a studentized range dist.

③ if $|\bar{R}_i - \bar{R}_j| > \text{critical range} \Rightarrow$

Reject $H_0: \overset{M}{M}d_i = M d_j$ and conclude that their popul^s have different medians. (i.e. they differ)

③ Ex cont'd

$H_0: m d_i = m d_j$ for $i, j = A, B, C$; $\alpha = .05$
 $i \neq j$

$k=3 \Rightarrow \binom{3}{2} = 3$, i.e. 3 comparisons

$$b=3$$

$$m_1=m_2=m_3$$

$$=4=b$$

$$\bar{R}_1 = \frac{TR_1}{m_1} = \frac{4}{4} = 1$$

$$\bar{R}_2 = \frac{TR_2}{m_2} = \frac{8}{4} = 2$$

$$\bar{R}_3 = \frac{TR_3}{m_3} = \frac{12}{4} = 3$$

$$= \sqrt{\frac{3(4)}{12(4)}} = .05$$

$$\text{and CR} = q_{\alpha}(K, \infty) \sqrt{\frac{K(K+1)}{12b}} = 1.655$$

$$q_{.05}(3, \infty) = 3.31$$

$$\Rightarrow |\bar{R}_1 - \bar{R}_2| = 1 \not> 1.655 \Rightarrow m_dA = m_dB$$

$$|\bar{R}_1 - \bar{R}_3| = 2 \geq 1.655 \Rightarrow m_dA \neq m_dC$$

$$|\bar{R}_2 - \bar{R}_3| = 1 \not> 1.655 \Rightarrow m_dB = m_dC$$

Notes

• Non-parametric tests are more powerful (greater proba of rejecting a false null hypothesis) than the F-test, if the assumptions for F-test violated.

• However, if F-test assumptions reasonably satisfied then F-test is more powerful & hence should be used.

i.e.

if the assumptions are satisfied then the F test is better @ detecting small differences than the parametric means than the non-parametric test.

- ① Fixed # of trials (n)
- ② trials are indep. of each other
- ③ trials result in k possible outcomes
(when $k=2 \Rightarrow$ Binomial distr.)
- ④ $P[\text{falling in category } i] = p_i$ remains constant from trial to trial, and $\sum_{i=1}^k p_i = 1$
i.e. laws of prob apply
- ⑤ let O_1, O_2, \dots, O_k are random variables, where $O_i =$ counts # of observ. in category i @ the end of n trials and $\sum_{i=1}^k O_i = n$

\Rightarrow if these 5 conditions are @ least reasonably satisfied, we have a multinomial distr.

- here we are interested in testing whether the proportions p_i in each category differ from a specific value or not.

i.e. $H_0: p_1 = p_{10}, p_2 = p_{20}, \dots, p_k = p_{k0}$) \propto
 $H_a: \text{@ least one } p_i \neq p_{i0}$

test stat $\chi^2 = \sum_{i=1}^k \frac{(O_i - E_i)^2}{E_i} \rightarrow$ "Pearson Chi-square test-statistic"

where $O_i =$ observed counts (i.e. observed # of observations in category i @ the end of n -trials)

$E_i =$ expected counts (i.e. # of observations in category i we would expect to have @ the end of n trials)

R.R. - We reject H_0 if $X^2 > X^2_{\alpha; (k-1)}$

Assumptions:

- ① A random sample taken from
- ② multinomial distr.
- ③ n is lrg enough for X^2 to apply (since X^2 is very sensitive to departures from normality we need $n > 30$ so that C.L.T. applies; i.e. we have approx. normality)

Ex1) Handout - Heating Gas Payments

	Accounts				
	Fully paid	1 mnth behind	2 mnth behind	> 2 mnth behind	Total
Observed	287	49	30	34	400
Expected	$np_{10} = 400(.8)$	$np_{20} = 400(.1)$	$np_{30} = 400(.06)$	$np_{40} = 400(.04)$	400
np_{i0}	= 320	= 40	= 24	= 16	400

$H_0: p_1 = .8, p_2 = .1, p_3 = .06, p_4 = .04; \alpha = .01$

$H_a: \text{at least one } p_i \neq p_{i0}$

Assume: ① random sample v (given)

② from multinomial dist.

③ n lrg ($n=400$) for X^2 to apply

$$\text{test stat. } X^2 = \sum_{i=1}^4 \frac{(O_i - E_i)^2}{E_i} = \frac{(287 - 320)^2}{320} + \frac{(49 - 40)^2}{40} + \frac{(30 - 24)^2}{24} + \frac{(34 - 16)^2}{16} = \underline{\underline{27.178}}$$

R.R. We reject H_0 if $X^2 > X^2_{\alpha; (k-1)} = X^2_{.01; (3)} = 11.34$

Since $X^2 = 27.178 > 11.34$, we reject H_0 and conclude that

① 1% significance level there is an evidence that this winters pattern of pay. is differs from the historical norm.

end. of email