

Split Plot Arrangement

The split plot arrangement is specifically suited for a two or more factor experiment.

This arrangement can be used with the CRD, RCBD, and LS designs discussed in this course.

Features of this design are that plots are divided into whole plots and subplots.

Example

Whole plots are wheat varieties (a_0 to a_3) and subplots are rates of a herbicide (b_0 to b_2).

a_1	a_2	a_3	a_0
b_1	b_2	b_0	b_2
b_2	b_0	b_1	b_0
b_0	b_1	b_2	b_1

With a split plot arrangement, the precision for the measurement of the effects of the whole plot factor(s) are sacrificed to improve that of the subplot factor.

Measurement of the subplot factor and its interaction with the main-plot factor is more precise than that obtained with an RCBD with a factorial arrangement.

Determining Which Factor to Use as the Whole and Subplot Factors

With the split plot arrangement, plot size and precision of measurement of the effects are not the same for whole and subplot factors. Thus, assignment of a particular factor to either the whole or subplot is extremely important. To make a choice, the following guidelines are suggested:

- 1. Degree of Precision:** for a greater deal of precision for factor B than factor A, assign factor B to the subplots and factor A to the whole plots.
- 2. Relative Size of the Main Effects:** If the main effect of one factor (e.g., factor A) is expected to be much larger and easier to detect than that of the other factor (e.g., factor B), factor A should be assigned to the whole plots and factor B to the subplots. This may increase the chances of detecting differences among levels of factor B.
- 3. Management Practices:** Cultural practices required by a factor may dictate use of large plots. In such a case, such factors should be assigned to whole plots.

Randomization and Layout

The randomization procedure for split plots consists of two parts:

1. Randomly assign whole plot treatments to whole plots based on the experimental design used.
2. Randomly assign subplot treatments to subplots. The randomization procedure has no effect on assignment of subplot treatments to subplots.

Expected Mean Squares for the Split Plot Arrangement

The example to be given will be for an RCBD with factors A and B considered as random effects.

Source of variation	df	Expected mean square
Replicate	$r-1$	$\sigma^2 + b\sigma_\gamma^2 + ab\sigma_R^2$
A	$a-1$	$\sigma^2 + b\sigma_\gamma^2 + r\sigma_{AB}^2 + rb\sigma_A^2$
Error (a) = Rep x A	$(r-1)(a-1)$	$\sigma^2 + b\sigma_\gamma^2$
B	$b-1$	$\sigma^2 + r\sigma_{AB}^2 + ra\sigma_B^2$
AxB	$(a-1)(b-1)$	$\sigma^2 + r\sigma_{AB}^2$
Error (b) = Rep x B(A)	$a(b-1)(r-1)$	σ^2
Total	$rab-1$	

ANOVA of a Split Plot Arrangement

Data Set

Treatments		Blocks (j)			Y _{i,k}
A _i	B _k	1	2	3	
a ₀	b ₀	13.8	13.5	13.2	40.5
	b ₁	15.5	15.0	15.2	45.7
	b ₂	21.0	22.7	22.3	66.0
	b ₃	18.9	18.3	19.6	56.8
Main plot total (Y _{0j})		69.2	69.5	70.3	209.0 = Y _{0..}
a ₁	b ₀	19.3	18.0	20.5	57.8
	b ₁	22.2	24.2	25.4	71.8
	b ₂	25.3	24.8	28.4	78.5
	b ₃	25.9	26.7	27.6	80.2
Main plot total (Y _{1j})		92.7	93.7	101.9	288.3 = Y _{1..}
Rep total (Y _j)		161.9	163.2	172.2	497.3 = Y _{...}

Treatment Totals Table

	a ₀	a ₁	ΣB _k
b ₀	40.5	57.8	98.3
b ₁	45.7	71.8	117.5
b ₂	66.0	78.5	144.5
b ₃	56.8	80.2	137.0
ΣA _i	209.0	288.3	497.3

Step 1. Calculate Correction Factor:

$$Y_{...}^2 / (r \times a \times b) = 497.3^2 / (3 \times 2 \times 4) = \mathbf{10,304.47}$$

Step 2. Calculate Total SS

$$\Sigma Y_{ijk}^2 - CF = (13.8^2 + 15.5^2 + 18.9^2 + \dots + 27.6^2) - CF = \mathbf{516.2}$$

Step 3. Calculate Replicate SS

$$\Sigma Y_{.j}^2 / axb - CF = (161.9^2 + 163.2^2 + 172.2^2) / (2 \times 4) - CF = \mathbf{7.87}$$

Step 4. Calculate A SS

$$\Sigma Y_{i.}^2 / rb - CF = (209^2 + 288.3^2) / (3 \times 4) - CF = \mathbf{262.02}$$

Step 5. Calculate Whole Plot SS

$$\Sigma Y_{ij}^2 / b - CF = (69.2^2 + 69.5^2 + \dots + 101.9^2) / 4 - CF = \mathbf{274.92}$$

Step 6. Calculate Whole Plot Error SS = Error(a) SS

$$\text{Whole Plot SS} - \text{A SS} - \text{Rep SS} = \mathbf{5.03}$$

Step 7. Calculate B SS

$$\Sigma Y_{.k}^2 / ra - CF = (98.3^2 + 117.5^2 + 144.5^2 + 137.0^2) / (3 \times 2) - CF = \mathbf{215.26}$$

Step 8. Calculate AxB SS

$$\begin{aligned} & \Sigma Y_{i.k}^2 / r - CF - \text{A SS} - \text{B SS} = (40.5^2 + 45.7^2 + \dots + 80.2^2) / 3 - CF - \text{A SS} - \text{B SS} \\ = & \mathbf{18.7} \end{aligned}$$

Step 9. Calculate Error(b) SS = Total SS - Rep SS - A SS - Error(a) SS - B SS - AxB SS

$$= \mathbf{7.24}$$

Step 10. Make ANOVA Table (Assuming A and B are fixed effects)

SOV	df	SS	MS	F
Replicate	2	7.87	3.935	6.53*
A	1	262.02	262.02	104.183**
Error(a)	2	5.03	2.515	
B	3	215.26	71.753	119.993**
AxB	3	18.70	6.233	10.337**
Error(b)	12	7.24	0.603	
Total	23	516.12		

LSD's for Split Plot Arrangement

1. To compare two whole plot means averaged over all subplot treatments (.e.g. a₀ vs. a₁)

$$t_{\alpha/2, err(a)df} \sqrt{\frac{2Error(a)MS}{rb}} = 4.303 \sqrt{\frac{2(2.515)}{3 \times 4}} = 2.79$$

2. To compare two subplot means average over all whole plot treatments (.e.g. b₀ vs. b₃)

$$t_{\alpha/2, err(b)df} \sqrt{\frac{2Error(b)MS}{ra}} = 2.179 \sqrt{\frac{2(0.603)}{3 \times 2}} = 0.98$$

3. To compare two subplot treatment means for the same whole plot treatment (.e.g. a₀b₀ vs. a₀b₃).

$$t_{\alpha/2, err(b)df} \sqrt{\frac{2Error(b)MS}{r}} = 2.179 \sqrt{\frac{2(0.603)}{3}} = 1.38$$

4. To compare two whole plot means at the same or different sub plot treatments (e.g. a_0b_0 vs. a_1b_0) or (e.g. a_0b_0 vs. a_1b_3).

$$t'_{\alpha/2,ab} \sqrt{\frac{2[(b-1)Error(b)MS + Error(a)MS]}{rb}}$$

Where t'_{ab} is a weighted estimate of t that can be calculated using the following formula:

$$t'_{\alpha/2,ab} = \frac{(b-1)Error(b)MS * t_{\alpha/2,Err(b)df} + Error(a)MS * t_{\alpha/2,Err(a)df}}{(b-1)Error(b)MS + Error(a)MS}$$

NOTE: This formula is to calculate t , not the degrees of freedom of t as was done for other situations.

$$t'_{ab} = \frac{(4-1)(.603)(2.179) + (2.515)(4.303)}{(4-1)(.603) + 2.515} = 3.414$$

$$\text{Therefore, the LSD} = 3.414 \sqrt{\frac{2[(4-1)(0.603) + 2.515]}{3 \times 4}} = 2.90$$

Table of Means for the example

B levels	A levels		Mean of B
	a_0	a_1	
b_0	13.5	19.3	16.4
b_1	15.2	23.9	19.6
b_2	22.0	26.2	24.1
b_3	18.9	26.7	22.8
Mean of A	17.4	24.0	

Table of Means for the example

				A levels			
B levels		a ₀		a ₁		Mean of B	
b ₀		13.5		19.3		16.4	
b ₁		15.2		23.9		19.6	
b ₂		22.0		26.2		24.1	
b ₃		18.9		26.7		22.8	
Mean of A		17.4		24.0			