

## Chapter 15

Experimental Group - Randomly selected to receive treatment

Control Group - Remaining selected to receive placebo

Comparative Experimental Study - randomization is employed to assign treatments

Treatments - defined by one or more explanatory factors

Comparative Observational Study - Random samples are taken from two or more populations and the observed outcomes are compared

Observational factors - levels of one or more explanatory factors that define populations

Mixed Experimental & Obs study - A two factor study consisting of both experimental and observational studies

Factor - an explanatory variable to be studied in an investigation

- experimental factor a factor where the level is assigned at random

- observational factor a factor that is characteristic of a experimental unit and not controlled by the investigator

- qualitative factor a categorical variable w/ r levels

- quantitative factor described by a numerical quantity

## Two-factor study (multi)

Combs	f1	f2
1	1	1
2	1	2
3	2	1
4	2	2
5	3	1
6	3	2

• Say factor 1 has 3 levels and factor 2 has 2 levels

• factor-level combos =  $3 \times 2 = 6$

• We can describe these experiments in the crossed table below

• If levels of one or more of the factors are unique to a particular level of another factor we

can use nesting.

- example 3 plants, 3 workers

from each plant, w/ the response production

over 5 batches

see table below

### Crossed

f2	f1		
	1	2	3
1	X	X	X
2	X	X	X

### Nested

Plant	Operator								
	1	2	3	4	5	6	7	8	9
1	X	X	X	X	X	X	X	X	X
2	X	X	X	X	X	X	X	X	X
3	X	X	X	X	X	X	X	X	X

\* Pick a subset of most important factors

\* Use a control when general effect of the treatment is unknown

\* Randomization - treatments assigned to experimental units at random

Blocking - Used to increase precision

- From one experiment create many

- Say we have  $Y_i = \beta_0 + \beta_1 X_i + \epsilon_i$

$$\sigma^2(b_1) = \frac{\sigma^2}{n}$$

- we can lower this by lowering  $\sigma^2$ , which we can do by blocking on a nuisance or confounding variable

say gender giving us:

$$Y_{if} = \beta_0 + \beta_1 X_{if} + \epsilon_{if}$$

$$Y_{im} = \beta_0 + \beta_1 X_{im} + \epsilon_{im}$$

Block	Randomization					
1	Treat	Treat	Placebo	---	Treatment	(male)
2	Placebo	Treat	Placebo	---	Treatment	(female)

$$Y_{ij} = \beta_0 + \beta_1 \underset{\substack{\uparrow \\ \text{treatment}}}{X_{ij1}} + \beta_2 \underset{\substack{\uparrow \\ \text{gender}}}{X_{ij2}} + \epsilon_{ij}$$

$\beta_0$  is mean response change due to treatment

$\beta_1$  is mean response change due to gender

NOTE:  $SSE(\text{Full Model}) = SSTO - SSR(X_1, X_2) = SSTO - [SSR(X_1) + SSR(X_2)]$

$$SSE(\text{Reduced}) = SSTO - SSR(X_1)$$

$$SSE(\text{Full}) = SSE(\text{Reduced}) - SSE(X_2)$$

∴  $SSE(X_2)$  extra sum of squares reduction when adding blocking

Completely Randomized Design (CRD) - Treatments are randomly assigned to experimental units

- most useful when experimental units are homogeneous

$$Y_{ij} = \beta_0 + \beta_1 \underset{\substack{\uparrow \\ \text{I(treat1)}}}{X_{ij1}} + \beta_2 \underset{\substack{\uparrow \\ \text{I(treat2)}}}{X_{ij2}} + \beta_3 \underset{\substack{\uparrow \\ \text{I(treat3)}}}{X_{ij3}} + \epsilon_{ij}$$

$\uparrow$  mean resp.

(64)

• We want  $\epsilon \sim N(0, \sigma^2) \rightarrow$  then we can use F test

Factorial Experiment - When treatments correspond to the set of all possible combinations of factor levels

$$Y_{ij} = \beta_0 + \beta_1 X_{ij1} + \beta_2 X_{ij2} + \beta_3 X_{ij1} X_{ij2} + \epsilon_{ij}$$

$\uparrow$  mean response     $\uparrow$  treat 1     $\uparrow$  treat 2     $\uparrow$  interaction     $\uparrow$  error

Randomized Complete Block Design - RCBD

• Experimental units are split into homogeneous blocks, then separate randomization of treatments for each block

Block	EXPER. UNIT			
	1	2	3	4
1	T <sub>1</sub>	T <sub>3</sub>	T <sub>4</sub>	T <sub>2</sub>
2	T <sub>3</sub>	T <sub>4</sub>	T <sub>1</sub>	T <sub>2</sub>

$$Y_{ij} = \beta_0 + \beta_1 X_{ij1} + \beta_2 X_{ij2} + \epsilon_{ij}$$

$\uparrow$  mean resp     $\uparrow$  treatment     $\uparrow$  blocking     $\uparrow$  error

• we want  $\epsilon \sim N(0, \sigma^2)$  then we can try F test

Nested Design: Blocking within a block

Block 1	Block 2	Exp Unit					
		1	2	3	4	5	6
1	1	X	X				
	2			X	X		
	3					X	X
2	1	X	X				
	2			X	X		
	3					X	X

Repeated Measures Design Each experimental unit receives all the treatments - like a taste test

Exper. Units	treatments		
	1	2	3
1			
2			
3			
4			
⋮			
12			

We can additionally block (Split Plot)

Block	Exp. Units	Treatments			
		1	2	3	
1	1				
	2				
	⋮				
	6				
	2	7			
		8			
9					
⋮					
12					

Incomplete Block Design When block sizes are smaller than treatments

Block	treatments				
	1	2	3	4	5
1	x	x	x		
2	x	x		x	
3	x	x			x
4	x		x	x	
5	x		x		x
6	x			x	x
7		x	x	x	
8		x	x		x
9		x		x	x
10			x	x	x

⊆ Balanced incomplete Block Design

15.4 and 15.5 omitted