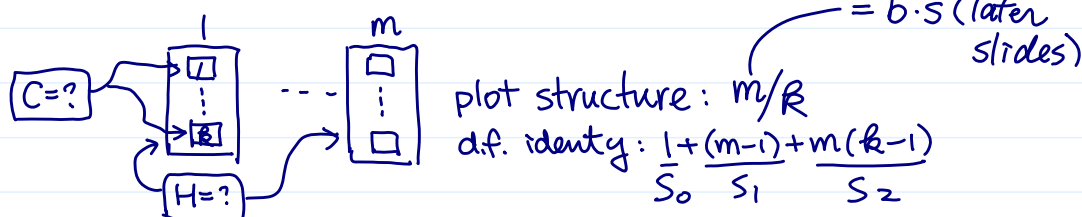


*treatment*  $\nabla$  *plot structure*

Stratum	source	df	SS	EMS	VR
	mean	mean	$\frac{\text{sum}^2}{N}$	$\ \tau_0\ ^2 + \xi_0$	—
$H$	large units	$n_H - 1$	$SS(H)$	$\frac{\ \tau_H\ ^2}{n_H - 1} + \xi_1$	$\frac{MS(H)}{MS(\text{large units residual})}$
	residual	—	—	$\xi_1$	—
	total	$m - 1$	$CSS(L) - \frac{\text{sum}^2}{N}$		
$C$	small units	$n_C - 1$	$SS(C)$	$\frac{\ \tau_C\ ^2}{n_C - 1} + \xi_2$	$\frac{MS(C)}{MS(\text{small units residual})}$
$H \wedge C$		$d_{HC}$	$SS(H \wedge C)$	$\frac{\ \tau_{HC}\ ^2}{d_{HC}} + \xi_2$	$\frac{MS(H \wedge C)}{MS(\text{small units residual})}$
	residual	—	—	$\xi_2$	—
	total	$m(k - 1)$	$\sum y_{\omega}^2 - CSS(L)$		
Total		$N$	$\sum y_{\omega}^2$		

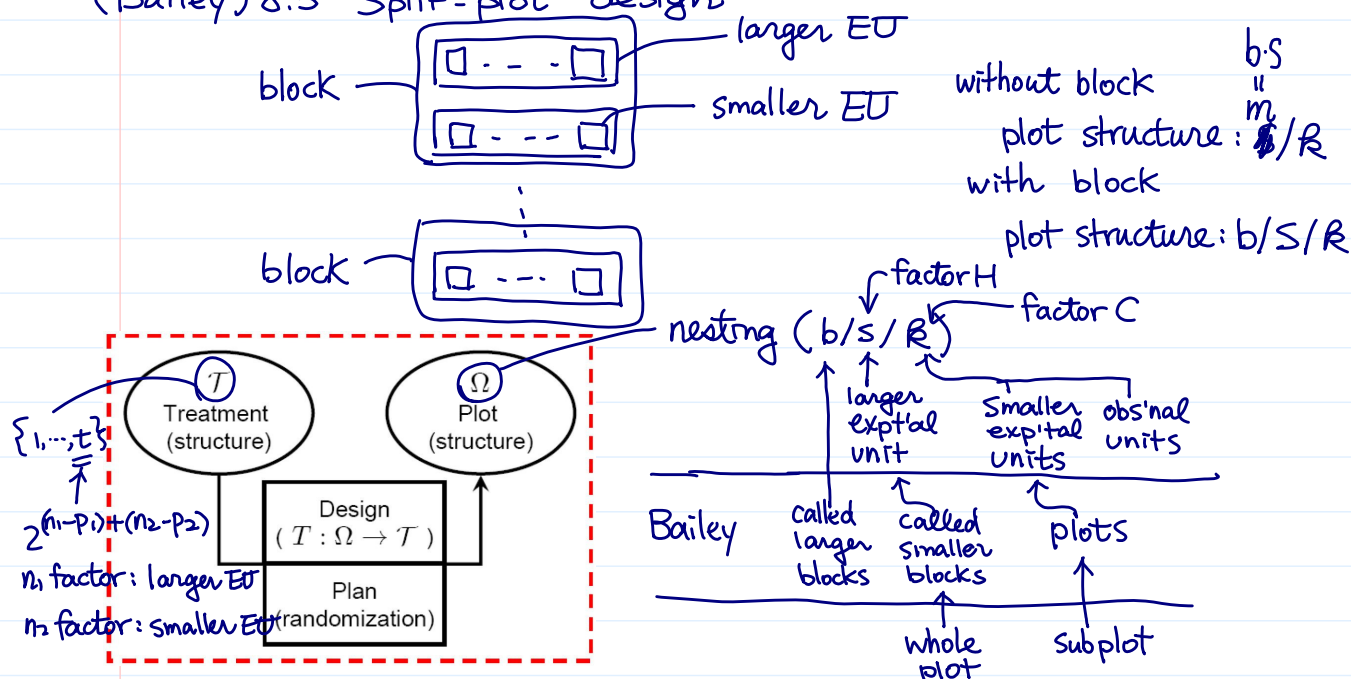
by subtraction



- usually, expect  $\xi_2 < \xi_1$  (c.f. ANOVA table in Lnp.7-3)  
 $\Rightarrow$  the design give more precision on the estimation & testing of main effects of  $C$   
 2-fi HC

7-6

## (Bailey) 8.3 Split-plot design



**Example 8.2 (Example 1.5 continued: Rye-grass)** Here the large blocks are the fields and the small blocks are the strips. Thus  $b = 2$ ,  $s = 3$  and  $k = 4$ .

**Example 8.3 (Animal-breeding)** In animal breeding it is typical to mate each *sire* (male parent) with several *dams* (female parents), each of which may then produce several offspring. Then the large blocks are the sires, the small blocks are the dams, and the plots are the offspring.

7-7

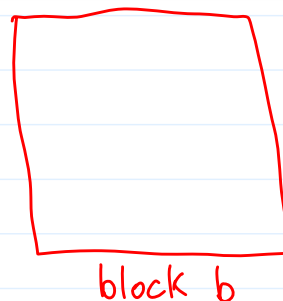
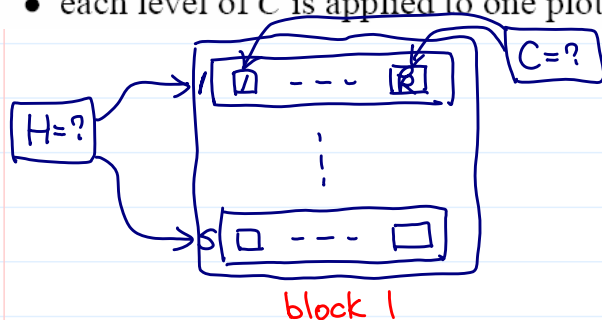
## Construction and randomization

restricted randomization

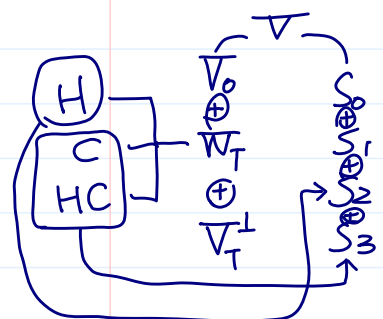
- (i) Apply levels of  $H$  to small blocks just as in a complete-block design.
- (ii) Within each small block independently, apply levels of  $C$  just as in a completely randomized design.

For simplicity, I shall describe only the classic *split-plot* design. This is like the second design in Section 8.2 except that

- the large units (small blocks) are grouped into  $b$  large blocks of size  $s$ ;
- each level of  $H$  is applied to one small block per large block (so  $n_H = s$  and  $r_H = b$ );
- each level of  $C$  is applied to one plot per small block (so  $n_C = k$  and  $r_C = 1$ ).



7-8



Stratum		df
$W_0$	mean	1
$W_B$	large blocks	$b-1$
$W_S$	small blocks	$b(s-1)$
$V_S^\perp$	plots	$bs(k-1)$
Total		$bsk$

plot structure

$b/s/R$

larger EU  
H

smaller EU  
C

d.f. identity

$$\frac{1+(b-1)+b(s-1)}{s_0} \frac{s_1^2 + bs(R-1)^2}{s_3}$$

Stratum	source	df	SS	EMS	VR
mean	mean	1	$\frac{\text{sum}^2}{bsk}$	$\ \tau_0\ ^2 + \xi_0$	—
large blocks	large blocks	$b-1$	$\text{CSS}(B) - \text{SS}(\text{mean})$	$\xi_B$	—
$H$	$H$	$s-1$	$\text{SS}(H)$	$\frac{\ \tau_H\ ^2}{s-1} + \xi_s$	$\frac{\text{MS}(H)}{\text{MS}(\text{small blocks residual})}$
residual	residual	$(b-1)(s-1)$	by subtraction	$\xi_s$	—
total	total	$b(s-1)$	$\text{CSS}(S) - \text{CSS}(B)$		
plots	$C$	$k-1$	$\text{SS}(C)$	$\frac{\ \tau_C\ ^2}{k-1} + \xi$	$\frac{\text{MS}(C)}{\text{MS}(\text{plots residual})}$
$H \wedge C$	$H \wedge C$	$(s-1)(k-1)$	$\text{SS}(H \wedge C)$	$\frac{\ \tau_{HC}\ ^2}{(s-1)(k-1)} + \xi$	$\frac{\text{MS}(H \wedge C)}{\text{MS}(\text{plots residual})}$
residual	residual	$(b-1)s(k-1)$	by subtraction	$\xi$	—
total	total	$bs(k-1)$	$\sum y_{\omega}^2 - \text{CSS}(S)$		
Total		$bsk$	$\sum y_{\omega}^2$		

(c.f. ANOVA table in LNp.7-6)

usually,  $\xi_H < \xi_s < \xi_B$

7-9

In general, plot structure :

plot factor  $V_1 \ V_2 \ V_3 \ \dots \ V_K$

treatment factors  $A_1, \dots, A_{t_1}$   
 $B_1, \dots, B_{t_2}$

$EU_A$   $EU_B$

$\hookrightarrow$  cause restriction on design:  $T: \Omega \rightarrow \mathcal{J}$

In the assignment of design key, i.e.,

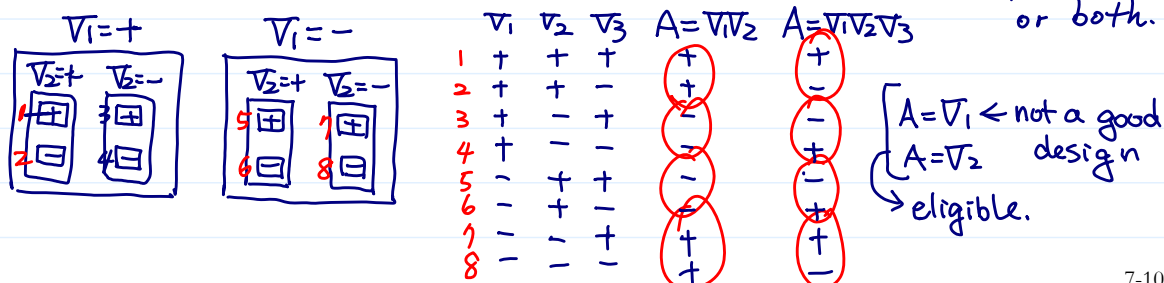
$$A_1 = V_1 \dots V_{t_1} \quad B_1 = V_{t_1+1} \dots V_{t_1+t_2}$$

$$\vdots \quad \vdots$$

$$A_{t_1} = V_1 \dots V_{t_1} \quad B_{t_2} = V_{t_1+1} \dots V_{t_1+t_2}$$

A cannot confound with plot effects involving  $V_3, V_4, \dots, V_K$

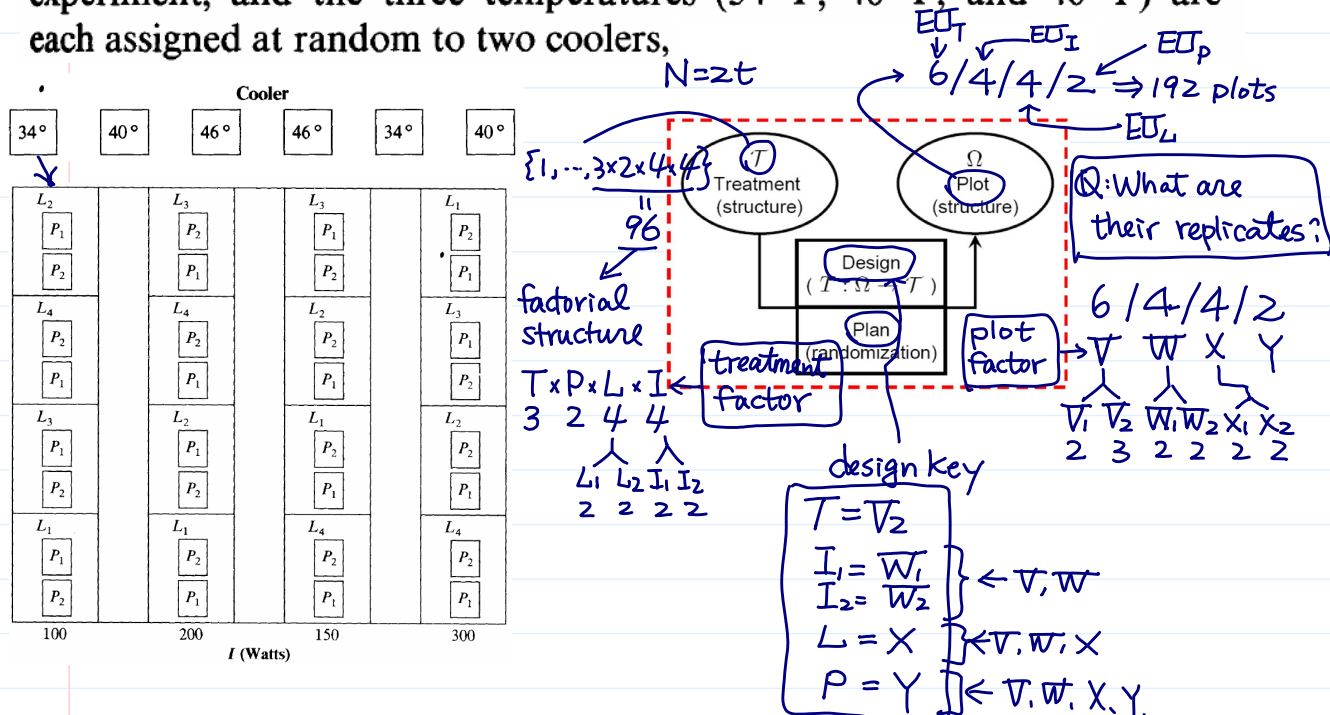
A must be confounded with plot effects formed by  $V_1$  or  $V_2$  or both.



7-10

(reading material 03)

A meat scientist wants to study the effect of temperature ( $T$ ) with three levels, types of packaging ( $P$ ) with two levels, types of lighting ( $L$ ) with four levels, and intensity of light ( $I$ ) with four levels on the color of meat stored in a meat cooler for seven days. Six coolers are available for the experiment, and the three temperatures (34°F, 40°F, and 46°F) are each assigned at random to two coolers,



7-11

	Source of Variation	df
$S_1$	Cooler Analysis	
	$T$	2
	Error(Cooler)	3
$S_2$	Column Analysis	
	Block	5
	$I$	3
	$I * T$	6
$S_3$	Partition Analysis	
	Block	23
	$L$	3
	$L * T$	6
	$L * I$	9
$S_4$	Half-Partition Analysis	
	Block	95
	$P$	1
	$P * T$	2
	$P * I$	3
	$P * I * T$	6
	$P * L$	3
	$P * L * T$	6
	$P * L * I$	9
	$P * L * I * T$	18
	Error(Half-Partition)	48

strata:

6/4/4/2

d.f. identity:

$$\frac{1}{S_0} + \frac{(6-1)}{S_1} + \frac{6 \cdot (4-1)}{S_2} + \frac{6 \cdot 4 \cdot (4-1)}{S_3} + \frac{6 \cdot 4 \cdot 4 \cdot (2-1)}{S_4}$$

$$T = V_2$$

$$I = W$$

$$L = X$$

$$P = Y$$

$$T = V_2$$

$$I = W$$

$$T I = V_2 W$$

$$L = X$$

$$L T = X V_2$$

$$L I = X W$$

$$L T I = X V_2 W$$

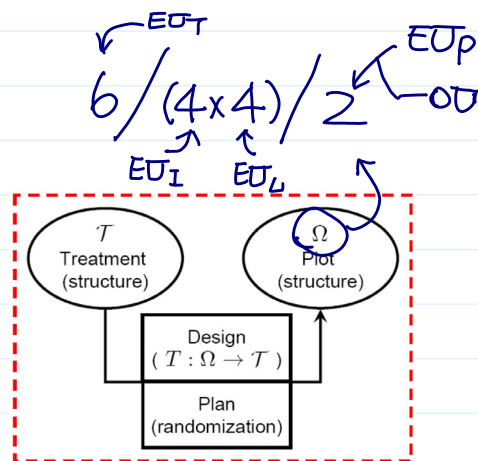
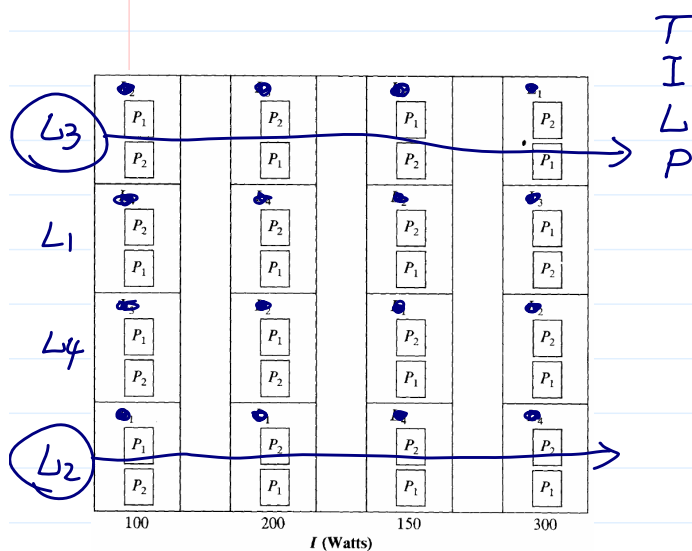
⋮

If  $T, W, X, Y$  are handled using fixed effect approach:

$T$  cannot be estimated.

$$S_2 \begin{pmatrix} 3 & -I \\ 6 & -I \times T \\ 3 & -W \\ 6 & \end{pmatrix}$$

7-12



$$6/(4 \times 4)/2$$

$$T \quad W \quad X \quad Y$$

d.f. identity

$$N(n_1, N(c(n_2, n_3), n_4))$$

On the choice of design key:

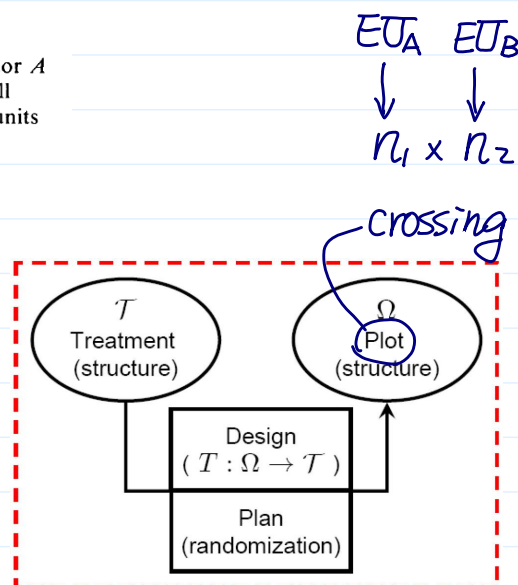
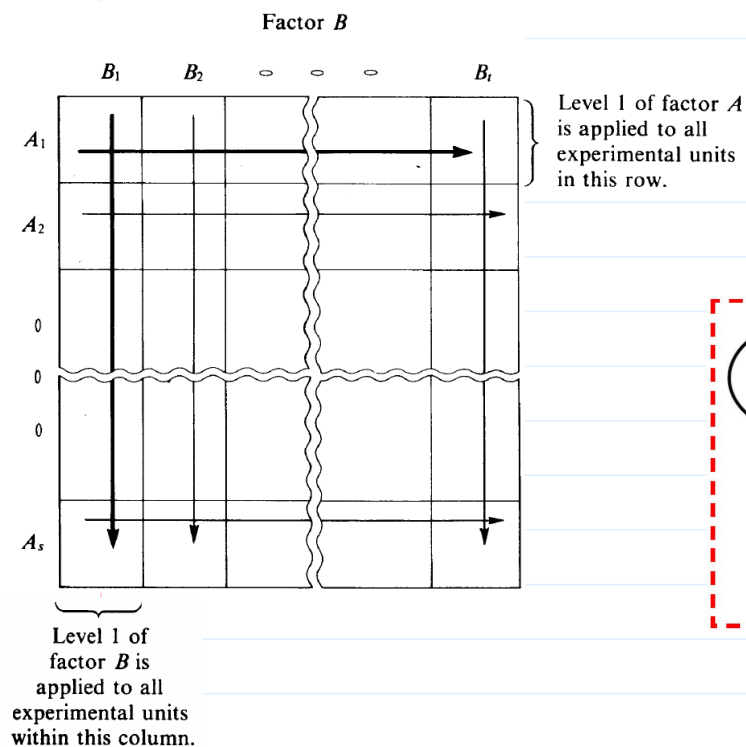
$I$  can be confounded with

plot effects involving  $T$  or  $W$  or  $TW$

$L : : : T$  or  $X$  or  $TX$

7-13

## (reading material 4) strip-plot design



7-14

Nitrogen Level		Irrigation Method	
		$I_1$	$I_2$
$N_1$	$N_1$	55	71
$N_3$	$N_3$	69	78
$N_2$	$N_2$	62	77

Nitrogen Level		Irrigation Method	
		$I_2$	$I_1$
$N_3$	$N_3$	81	77
$N_1$	$N_1$	77	63
$N_2$	$N_2$	79	66

Nitrogen Level		Irrigation Method	
		$I_1$	$I_2$
$N_2$	$N_2$	70	78
$N_3$	$N_3$	79	80
$N_1$	$N_1$	63	77

Nitrogen Level		Irrigation Method	
		$I_1$	$I_2$
$N_3$	$N_3$	76	79
$N_2$	$N_2$	66	76
$N_1$	$N_1$	65	75

d.f. identity

$$N(n_1, c(n_2, n_3))$$

$$= 1 + (n_1 - 1) + n_1(c(n_2, n_3) - 1)$$

$$1 + (n_1 - 1) + n_1((n_2 - 1) + (n_3 - 1) + (n_2 - 1)(n_3 - 1))$$

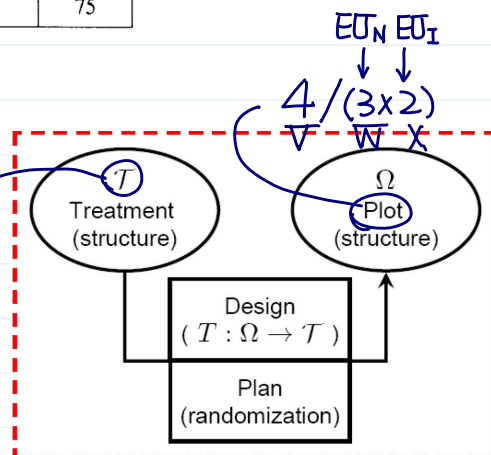
$$= 1 + \frac{(n_1 - 1)}{s_0} + n_1 \frac{(n_2 - 1)}{s_1} + n_1 \frac{(n_3 - 1)}{s_2} + n_1 \frac{(n_2 - 1)(n_3 - 1)}{s_4}$$

Source of Variation	df
$S_0$ 1 Replication or Set	③
$S_1$ 3 Error(Row)	2
$S_2$ 8 Error(Column)	6
$S_3$ 4 Error(Unit)	1
$S_4$ 8	3
	2
	6

$\{1, \dots, 3 \times 2\}$

factorial structure  $N \times I$

3 2



$$\begin{matrix} N=W \\ I=X \\ NI=WX \end{matrix}$$

design key

7-15