

# Paired Comparison Design (Sewage Experiment)

- Objective :** To compare two methods MSI and SIB for determining chlorine content in sewage effluents;  $y =$  residual chlorine reading.

\* response : reading  
 \* treatment factor : method (qualitative)  
 2 levels - MSI, SIB

Table 1: Residual Chlorine Readings, Sewage Experiment

$\alpha_i$ 's also removed in  $\bar{y}_{SIB} - \bar{y}_{MSI}$ . ( $= \bar{d}$ ) but  $\alpha_i$ 's still influence  $\sigma^2$

Sample	Method		$d_i$
	MSI	SIB	
1	0.39	0.36	-0.03
2	0.84	1.35	0.51
3	1.76	2.56	0.80
4	3.35	3.92	0.57
5	4.69	5.35	0.66
6	7.70	8.33	0.63
7	10.52	10.70	0.18
8	10.92	10.91	-0.01

blocks

$y_{SIB,i} - y_{MSI,i}$  (within block comparison)

Successful blocking  
 btwn-block variation: large  
 within-block variation: small

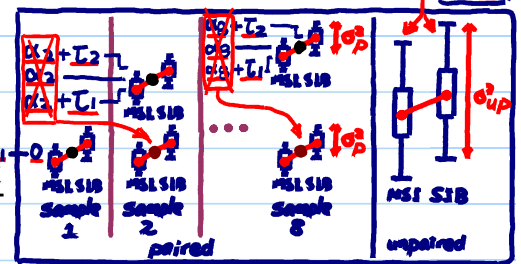
\* block factor  
 Sample (qualitative)  
 8 levels - 1, 2, ..., 8  
 block size = 2

\* Exp'tal unit = ? 16 EUs  
 \* Design matrix

method	sample	y
MSI	1	0.39
SIB	1	0.36
MSI	2	0.84
SIB	2	1.35
...	...	...
MSI	8	10.92
SIB	8	10.91

Data matrix

- Experimental Design :** Eight samples were collected at different doses and contact times. Two methods were applied to each of the eight samples. It is a paired comparison design because the pair of treatments are applied to the same samples (or units).



# Paired Comparison Design vs. Unpaired Design

one-sample problem:  $Z_1, \dots, Z_n$  iid  $\sim N(\mu, \sigma^2)$ ,  $H_0: \mu = \mu_0$  a known constant

**Paired Comparison Design :** Two treatments are randomly assigned to each block of two units. Can eliminate block-to-block variation and is effective if such variation is large. Examples: pairs of twins, eyes, kidneys, left and right feet. (Subject-to-subject variation much larger than within-subject variation).



btwn block variation

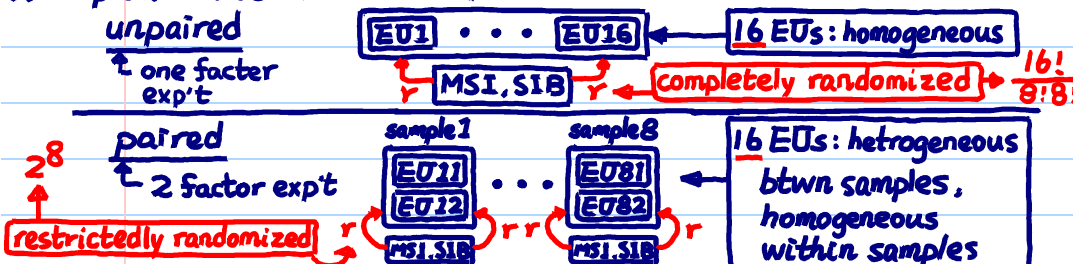
For block effects, we are not interested in their estimation or testing. But, if we ignore block effects (either before or after exp't), the analysis of treatment effects suffer.

$$Y = X_1 B_1 + X_2 B_2 + \epsilon \rightarrow X_1 B_1 + \cancel{X_2 B_2} + (I - H_1) X_2 B_2 + \epsilon^*$$

- Unpaired Design :** Each treatment is applied to a separate set of units, or called the two-sample problem. Useful if pairing is unnecessary; also it has more degrees of freedom for error estimation (see LNp. 4-4).

btwn block variation  $\approx 0$   
 $\Rightarrow$  homogeneous EUs

\* Exp'tal units: water samples




2-sample problem: 1-way layout of a 2-level factor  
 $y_{1j}$ 's iid  $\sim N(\mu_1, \sigma^2)$   
 $y_{2j}$ 's iid  $\sim N(\mu_2, \sigma^2)$   
 $H_0: \mu_1 = \mu_2$

# Paired $t$ tests ← conceptual model: $y \sim \beta_0 + \text{sample} + \text{method} + \epsilon \dots (*)$

- Paired  $t$  test: Let  $y_{i1}, y_{i2}$  be the responses of treatments 1 and 2 for unit  $i, i = 1, \dots, N$ . Let  $d_i = y_{i2} - y_{i1}, \bar{d}$  and  $s_d^2$  the sample mean and variance of  $d_i$ .

7 parameter → (main-effect model, no interaction btwn block & treatment factors, why?)

one-sample problem  
 $d_1, \dots, d_N \text{ iid } \sim N(\mu, \sigma^2)$   
 $H_0: \mu = 0$   
  
 $\text{Var}(\bar{d}) = \sigma^2 / N$   
 $t_{paired} = \frac{\bar{d} - 0}{s_d / \sqrt{N}}$   
 $S_d^2 = \sum (d_i - \bar{d})^2 / (N-1)$

# of blocks  $E\epsilon_i \rightarrow \sum \epsilon_i \rightarrow y$   
 F-test for "method"  
 $\Omega: (*)$   
 $\omega: \beta_0 + \text{sample} + \epsilon$   
 both models contain "sample". The variation caused by "sample" is removed from  $\epsilon$ .  
 Always put block factors before treatment factors in sequential ANOVA (why?)

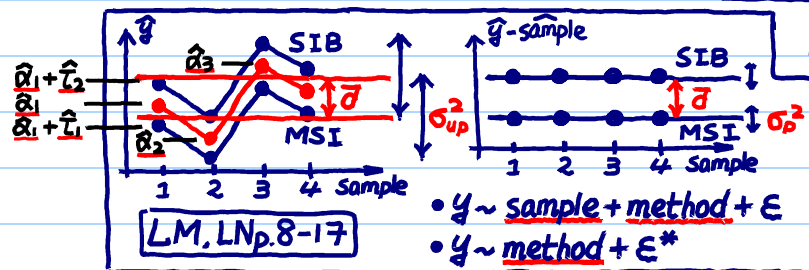
The two treatments are declared significantly different at level  $\alpha$  if

$y_{ij} = \beta_0 + \beta_1 + \beta_2 + \epsilon_{ij}$   
 $\hat{\epsilon}_{i1} = -\hat{\epsilon}_{i2}$   
 $d_i = \hat{\epsilon}_{i2} - \hat{\epsilon}_{i1} = 2\hat{\epsilon}_{i2}$   
 $\bar{d} = 2\bar{\epsilon}_{i2}$   
 $d_i - \bar{d} = \hat{\epsilon}_{i2} - \bar{\epsilon}_{i2} = 2\hat{\epsilon}_{i2}$

$|t_{paired}| > t_{N-1, \alpha/2}$

$(t_{paired})^2 = F = \frac{\|P_{\Omega}(\bar{y})\|^2 / 1}{RSS_{\Omega} / df_{\Omega}}$  (1)

total df:  $2N$   
 df for intercept: 1  
 df for treatment factor: 1  
 df for block factor:  $N-1$   
 df for residuals:  $2N - 1 - 1 - (N-1) = N-1$



$\Omega \Theta \omega = \text{span}\{y\}$   
 $y = \begin{bmatrix} -1 \\ \vdots \\ -1 \end{bmatrix} \begin{matrix} y_{i1}'s \\ y_{i2}'s \end{matrix}$

# Unpaired $t$ tests ← (one-way layout, 2-sample problem) conceptual model: $y \sim \beta_0 + \text{method} + \epsilon' \dots (\Delta)$

- Unpaired  $t$  test: The unpaired  $t$  test is appropriate if we randomly choose  $N$  of the  $2N$  units to receive one treatment and assign the remaining  $N$  units to the second treatment. Let  $\bar{y}_i$  and  $s_i^2$  be the sample mean and sample variance for the  $i$ th treatment,  $i = 1$  and  $2$ . Define

1 parameter

F-test for method

$\Omega: (\Delta)$

$\omega: y = \beta_0 + \epsilon'$

$F = \frac{\|P_{\Omega}(\bar{y})\|^2 / 1}{RSS_{\Omega} / df_{\Omega}}$

$t_{unpaired} = \frac{(\bar{y}_2 - \bar{y}_1) / \sqrt{(s_2^2/N) + (s_1^2/N)}}{s_p / \sqrt{2}}$   
 $\text{Var}(\bar{y}_2 - \bar{y}_1) = \sigma_2^2/N + \sigma_1^2/N$

(exercise)

total df:  $2N$   
 df for intercept: 1  
 df for treatment factor: 1  
 df for residuals:  $2N - 1 - 1 = 2N - 2$  (2)

The two treatments are declared significantly different at level  $\alpha$  if

$|t_{unpaired}| > t_{2N-2, \alpha/2}$

- Note that the degrees of freedom in (1) and (2) are  $N-1$  and  $2N-2$  respectively. Unpaired  $t$ -test has more df's, but make sure that the unit-to-unit variation is under control (if this method is to be used).

$t_{2N-2, \alpha/2} < t_{N-1, \alpha/2}$

If block effects significant (large btwn block variation) model (\*) is better  $\Rightarrow \hat{\sigma}_{(\Delta)}^2 \gg \hat{\sigma}_{(*)}^2$

If block effects insignificant (btwn block variation  $\approx 0$ ) model ( $\Delta$ ) is better ( $\Rightarrow \hat{\sigma}_{(\Delta)}^2 \approx \hat{\sigma}_{(*)}^2$ )  $\Rightarrow t_{unpaired}$  has high power, and we can get better (accurate) estimate of  $\sigma^2$  under ( $\Delta$ )