

In this Chapter: Model selection  
as a result of choices for  $\sum$

23. Repeated Measures  $\text{Var}(y)$

# Repeated Measures Studies

- Measure the experimental unit multiple times
- Measurements often taken over time
- Observations over time typically correlated
- Longitudinal versus crossover study designs
- Repeated measures analysis accounts for correlation

pay attention to  
time points being  
equally  
spaced vs. not

## Repeated Measures Example

In an exercise therapy study, subjects were assigned to one of three weightlifting programs

- i=1: The number of repetitions of weightlifting was increased as subjects became stronger.
- i=2: The amount of weight was increased as subjects became stronger.
- i=3: Subjects did not participate in weightlifting.

- Measurements of strength ( $y$ ) were taken on days 2, 4, 6, 8, 10, 12, and 14 for each subject.
- Source: Littel, Freund, and Spector (1991), SAS System for Linear Models.
- R code: RepeatedMeasures.R
- SAS code: RepeatedMeasures.sas

Program 1

$S_{1I}$
2 4 6 8 10 12 14

•  
•  
•

$S_{1n_1}$

2	4	6	8	10	12	14
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$n_1$

Program 2

$S_{2I}$
2 4 6 8 10 12 14

•  
•  
•

$S_{2n_2}$

2	4	6	8	10	12	14
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$n_2$

Program 3

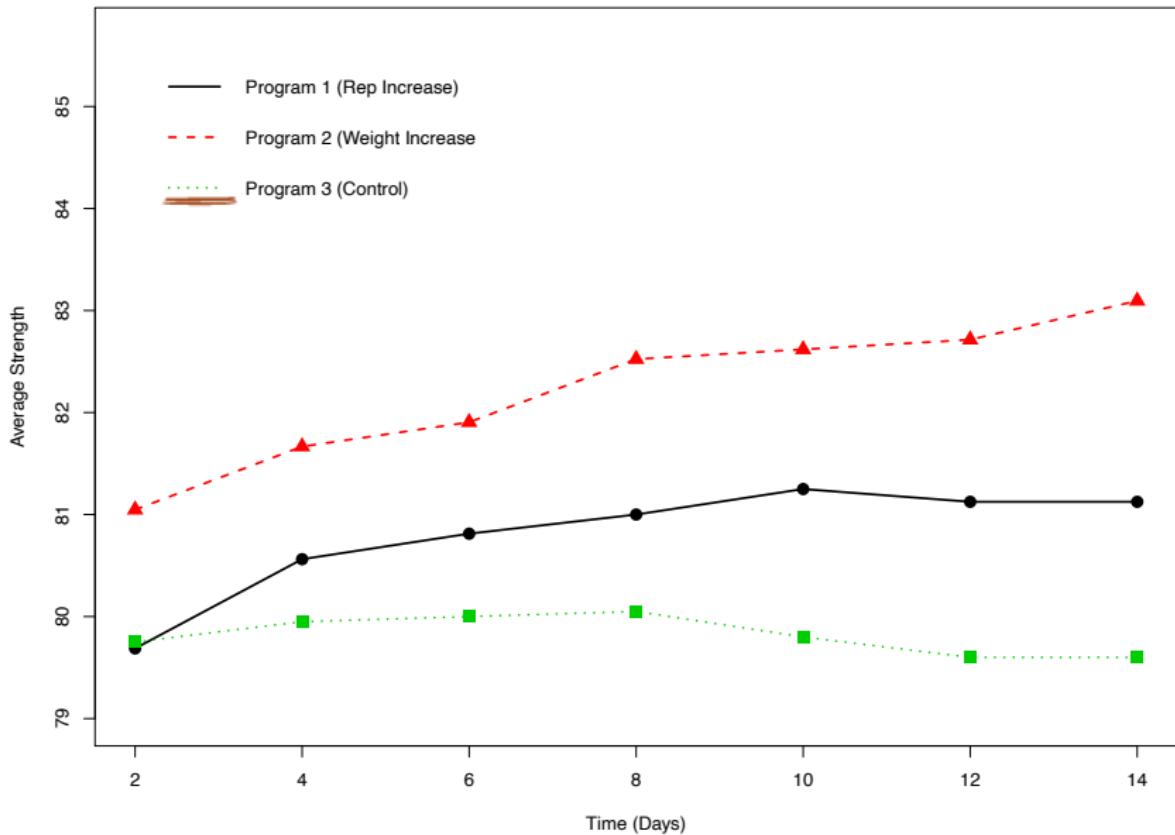
$S_{3I}$
2 4 6 8 10 12 14

•  
•  
•

$S_{3n_3}$

2	4	6	8	10	12	14
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$n_3$



# A Linear Mixed-Effects Model

Let  $y_{ijk}$  be the strength measurement for program  $i$ , subject  $j$ , and time point  $k$ . Suppose

$$y_{ijk} = \underline{\mu} + \underline{\alpha_i} + s_{ij} + \underline{\tau_k} + \underline{\gamma_{ik}} + e_{ijk},$$

*fixed effects*

*random effects*

where  $\mu, \alpha_1, \alpha_2, \alpha_3, \tau_1, \dots, \tau_7$ , and  $\gamma_{11}, \dots, \gamma_{37}$  are unknown real-valued parameters, and

$$s_{ij} \stackrel{iid}{\sim} \mathcal{N}(0, \sigma_s^2) \text{ independent of } e_{ijk} \stackrel{iid}{\sim} \mathcal{N}(0, \sigma_e^2).$$

Note that this model is the same model we would use for a split-plot experiment in which the whole-plot part of the experiment has a completely randomized design.

Subjects are the whole-plot experimental units, and measurement occasions within subject are treated like split-plot experimental units.

## ANOVA Table

Source	DF
Program $\frac{k}{t}$ fixed	$3 - 1$
Subject(Program) ran	$(16 - 1) + (21 - 1) + (20 - 1)$
Time $\frac{m}{t}$ fixed	$m_1 \quad m_2 \quad m_3$ $7 - 1$
Program $\times$ Time fixed	$(3 - 1)(7 - 1)$
Time $\times$ Subject(Program) ran	$(7 - 1)(57 - 3)$
C. Total	$57 \times 7 - 1$

The measurement occasions are not really split-plot experimental units because levels of the factor time (2, 4, ..., 14) were not randomly assigned to measurement occasions.

Nonetheless, this split-plot model might be reasonable for some experiments where experimental units are measured repeatedly over time.

Average strength after  $2k$  days on the  $i$ th program is

$$\begin{aligned} E(\underline{y_{ijk}}) &= E(\underline{\mu} + \underline{\alpha_i} + \underline{s_{ij}} + \underline{\tau_k} + \underline{\gamma_{ik}} + \underline{e_{ijk}}) \\ &= \mu + \alpha_i + \underbrace{E(s_{ij})}_{=0} + \tau_k + \gamma_{ik} + \underbrace{E(e_{ijk})}_{=0} \\ &= \mu + \alpha_i + \tau_k + \gamma_{ik} \end{aligned}$$

for  $i = 1, 2, 3$  and  $k = 1, 2, \dots, 7$ .

= Mean structure

The variance of any single observation is

$$\boxed{\text{Var}(y_{ijk})} = \text{Var}(\mu + \alpha_i + s_{ij} + \tau_k + \gamma_{ik} + e_{ijk}) \\ = \text{Var}(s_{ij} + e_{ijk})$$

*indep.*  $\equiv \text{Var}(s_{ij}) + \text{Var}(e_{ijk})$

$$= \underline{\sigma_s^2} + \underline{\sigma_e^2}.$$

The covariance between any two different observations from the same subject is

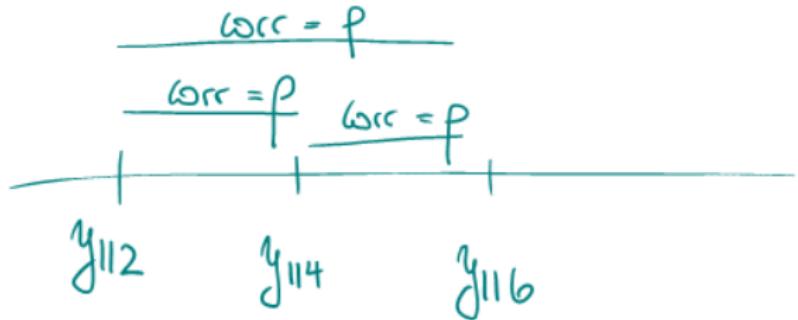
$$\begin{aligned}\text{Cov}(y_{ijk}, y_{ijl}) &= \text{Cov}(\mu + \alpha_i + s_{ij} + \tau_k + \gamma_{ik} + e_{ijk}, \\ &\quad \mu + \alpha_i + s_{ij} + \tau_\ell + \gamma_{il} + e_{ijl}) \\ &= \text{Cov}(s_{ij} + e_{ijk}, s_{ij} + e_{ijl}) \\ &\quad = 0 \\ &= \text{Cov}(s_{ij}, s_{ij}) + \text{Cov}(s_{ij}, e_{ijl}) \\ &\quad + \text{Cov}(e_{ijk}, s_{ij}) + \text{Cov}(e_{ijk}, e_{ijl}) \\ &\quad = 0 \\ &= \text{Var}(s_{ij}) = \sigma_s^2.\end{aligned}$$

The correlation between  $y_{ijk}$  and  $y_{ijl}$  is

$$\frac{\sigma_s^2}{\sigma_s^2 + \sigma_e^2} \equiv \rho.$$

regardless of time  
point we assume  
the same correlation:

Observations taken on different subjects are uncorrelated.



matrix for 1 individual (out of the 57)

For the set of observations taken on a single subject, we have

$$\text{Var} \left( \begin{bmatrix} y_{ij1} \\ y_{ij2} \\ \vdots \\ y_{ij7} \end{bmatrix} \right) = \begin{bmatrix} \sigma_e^2 + \sigma_s^2 & \sigma_s^2 & \cdots & \sigma_s^2 \\ \sigma_s^2 & \sigma_e^2 + \sigma_s^2 & \cdots & \sigma_s^2 \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_s^2 & \sigma_s^2 & \sigma_s^2 & \sigma_e^2 + \sigma_s^2 \end{bmatrix}$$
$$= \sigma_e^2 \mathbf{I}_{7 \times 7} + \sigma_s^2 \mathbf{1}\mathbf{1}_{7 \times 7}^\top.$$

This is known as a compound symmetric covariance structure.

$\sigma_e^2$  &  $\sigma_s^2$  requiring estimation

Using  $n_i$  to denote the number of subjects in the  $i$ th program, we can write this model in the form

$$\mathbf{y} = \underline{\mathbf{X}\boldsymbol{\beta}} + \underline{\mathbf{Z}\mathbf{u}} + \underline{\mathbf{e}}.$$

To make things slightly easier to write, let

$$\mathbf{y}_{ij} = [y_{ij1}, y_{ij2}, y_{ij3}, y_{ij4}, y_{ij5}, y_{ij6}, y_{ij7}]^\top$$

and

$$\mathbf{e}_{ij} = [e_{ij1}, e_{ij2}, e_{ij3}, e_{ij4}, e_{ij5}, e_{ij6}, e_{ij7}]^\top$$

for all  $i = 1, 2, 3$  and all  $j = 1, \dots, n_i$ .

$$\begin{bmatrix}
 y_{11} \\
 y_{12} \\
 \vdots \\
 y_{1n_1} \\
 \hline
 y_{21} \\
 y_{22} \\
 \vdots \\
 y_{2n_2} \\
 \hline
 y_{31} \\
 y_{32} \\
 \vdots \\
 y_{3n_3}
 \end{bmatrix}
 = M_{kt} \begin{bmatrix}
 1_{7 \times 1} & 1_{7 \times 1} & 0_{7 \times 1} & 0_{7 \times 1} \\
 1_{7 \times 1} & 1_{7 \times 1} & 0_{7 \times 1} & 0_{7 \times 1} \\
 \vdots & \vdots & \vdots & \vdots \\
 1_{7 \times 1} & 1_{7 \times 1} & 0_{7 \times 1} & 0_{7 \times 1} \\
 1_{7 \times 1} & 0_{7 \times 1} & 1_{7 \times 1} & 0_{7 \times 1} \\
 1_{7 \times 1} & 0_{7 \times 1} & 1_{7 \times 1} & 0_{7 \times 1} \\
 \vdots & \vdots & \vdots & \vdots \\
 1_{7 \times 1} & 0_{7 \times 1} & 1_{7 \times 1} & 0_{7 \times 1} \\
 1_{7 \times 1} & 0_{7 \times 1} & 0_{7 \times 1} & 1_{7 \times 1} \\
 1_{7 \times 1} & 0_{7 \times 1} & 0_{7 \times 1} & 1_{7 \times 1} \\
 \vdots & \vdots & \vdots & \vdots \\
 1_{7 \times 1} & 0_{7 \times 1} & 0_{7 \times 1} & 1_{7 \times 1}
 \end{bmatrix}
 \begin{array}{c}
 \text{control} \\
 \times
 \end{array}
 \begin{array}{c}
 \text{time} \\
 \tilde{\gamma}_1 \dots \tilde{\gamma}_7
 \end{array}
 \begin{bmatrix}
 \beta \\
 \mu \\
 \alpha_1 \\
 \alpha_2 \\
 \alpha_3 \\
 \hline
 \tau_1 \\
 \tau_2 \\
 \tau_3 \\
 \tau_4 \\
 \tau_5 \\
 \tau_6 \\
 \tau_7 \\
 \hline
 \gamma_{11} \\
 \gamma_{12} \\
 \vdots \\
 \gamma_{37}
 \end{bmatrix}$$

interaction  
btw.  $\alpha_3$  & time 7

# End lecture 37 +1 = Lecture 38

04-30-25

$$[I_{(n_1+n_2+n_3) \times (n_1+n_2+n_3)} \otimes \mathbf{1}_{7 \times 1}]$$

2

$$\begin{bmatrix} s_{11} \\ s_{12} \\ \vdots \\ s_{1n_1} \\ s_{21} \\ s_{22} \\ \vdots \\ s_{2n_2} \\ s_{31} \\ s_{32} \\ \vdots \\ s_{3n_3} \end{bmatrix} + \begin{bmatrix} e_{11} \\ e_{12} \\ \vdots \\ e_{1n_1} \\ e_{21} \\ e_{22} \\ \vdots \\ e_{2n_2} \\ e_{31} \\ e_{32} \\ \vdots \\ e_{3n_3} \end{bmatrix}$$

= k e