Mixed model analysis case studies

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Mixed model analysis of randomized studies:

 evaluation of new mathematics teaching method in primary school

comparison of whole grain vs. refined grain diets

TRACK (Teaching Routines and Content Knowledge) project

123 Danish primary schools randomized into treatment and control group.

Treatment: new mathematics teaching method inspired by Singaporean practice.

Follow pupils over three years starting with 4th grade.

Data available now: mathematics test result at beginning of study (baseline).

My colleagues from Aarhus wanted to test adequate randomization by assessing treatment effect at baseline.

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Is this a good idea ?

They used ordinary least squares (OLS) but with adjusted standard errors taking into account correlation between schools and classes.

$$\hat{\beta} = (X^{\mathsf{T}}X)^{-1}X^{\mathsf{T}}Y \quad \mathbb{V}\mathrm{ar}\hat{\beta} = (X^{\mathsf{T}}X)^{-1}X^{\mathsf{T}}VX(X^{\mathsf{T}}X)^{-1}$$

Plug-in empirical estimate of V.

Note: this is not BLUE !

They found slightly significant negative treatment effect !! (*p* slightly less than 5%)

Conclusion ??

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I used linear mixed models with school and class random effects.

In this case

$$\hat{eta} = (X^{\mathsf{T}}V^{-1}X)^{-1}X^{\mathsf{T}}V^{-1}Y \quad \mathbb{V}\mathrm{ar}\hat{eta} = (X^{\mathsf{T}}V^{-1}X)^{-1}$$

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Is this BLUE ?

Possible disadvantage of mixed model analysis ?

Results regarding fraction (brøk-regning) test results

Estimates of treatment effects and *p*-values:

OLS	OLS <i>p</i> -value	permutation <i>p</i> -value
-2.39	0	0.014
Mixed (BLUE)	approx <i>t p</i> -value	permutation
-1.34	0.151	0.145

Permutation: randomly permute schools into treatment and control and assess treatment effect for each permuted data set. Under null-hypothesis, data should come from this permutation distribution.

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Whole grain (WG) vs. refined grain (RG)

Subjects randomly allocated to two treatment arms:

Group 1: baseline WG RG Group 2: baseline RG WG

Outcome: LDL cholesterol in blood

Note: possible cross over effect (treatment effect WG-RG may depend on order of treament (WG first or last)

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For *i*th subject three measurements Y_{it} , t = 1, 2, 3

Standard approach: regression using baseline Y_{1t} as covariate:

$$Y_{it} = \mu_{it} + \alpha Y_{i1} + \epsilon_{it}, \quad t = 2, 3$$

 μ_{it} : two sided ANOVA based on Group (1, 2) and Treatment (WG, RG)

Problem: we need to skip all observations for i if baseline is missing !

Alternative: mixed model with subject random effect

$$Y_{it} = \mu_{it} + U_{it} + \epsilon_{it}, \quad t = 1, 2, 3$$

Specification of μ_{it} more complicated since Treatment now has three levels WG, RG and baseline. Due to randomization, no group effect for baseline !

Results: no cross over effect, WG good for reducing LDL :)