

Linear Mixed Models

Introduction

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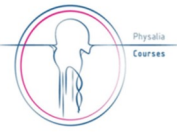
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Our questions



If we want to answer the following questions....



Time effect: What is the shape of the trajectory of the mean response over time?

Group effect: What is the average difference between groups of individuals?

Interaction between time and group: How does the relationship between the response and time vary according to groups of individuals?

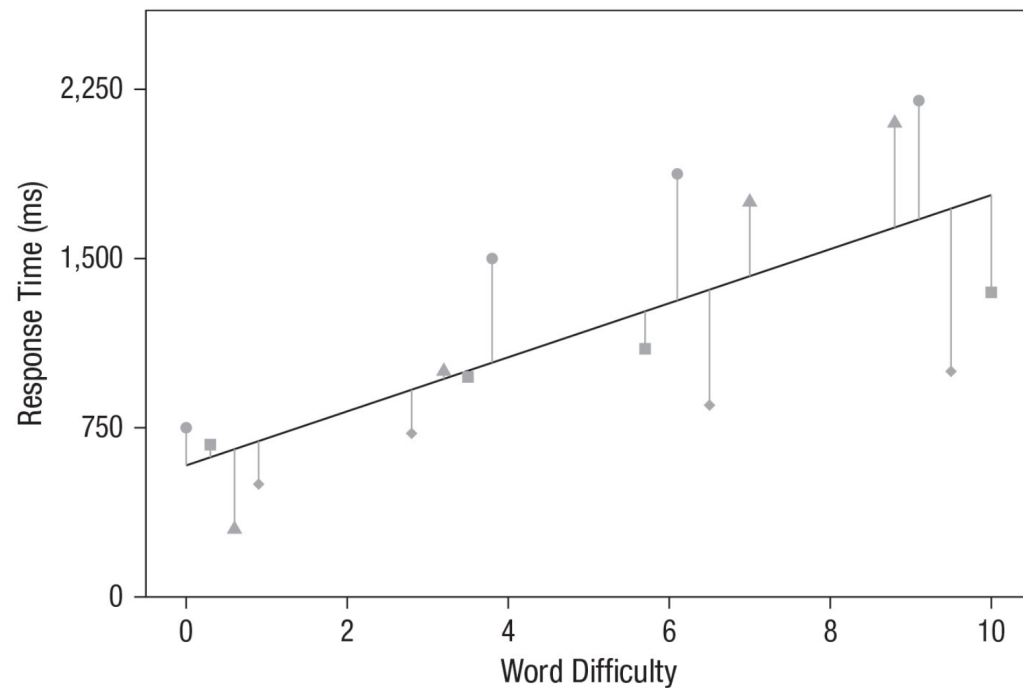


Linear Mixed Models

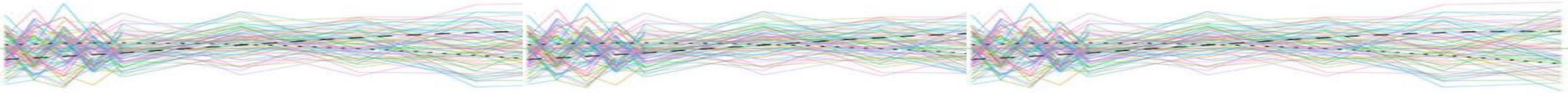
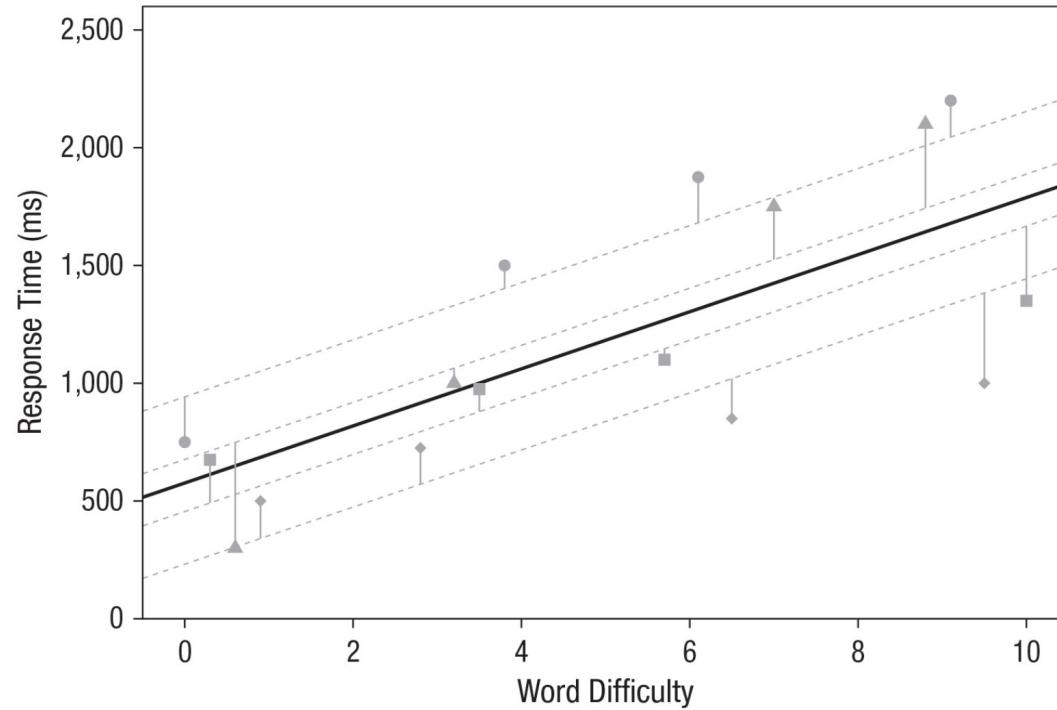
- Natural heterogeneity across subjects → Some subset of the regression coefficients vary randomly from one individual to another
- Individuals in population are assumed to have their own subject-specific mean response trajectories over time.



Fixed effects only



Mixed – fixed + random



Fixed and Random effects

The mean response is modeled as a combination of population characteristics (***fixed effects***) assumed to be shared by all individuals, while subject-specific effects (***random effects***) are unique to a particular individual.



Mixed Models

Linear Mixed Models are a particular type of hierarchical models which contain both fixed and random effects.

one-way ANOVA we have a single 'treatment' factor with several levels (= **groups**), and replicated observations at each level.



Empty model

one-way random-effect ANOVA
model, is $Y_{ij} = \beta_0 + u_{0i} + \epsilon_{ij}$

where β_0 is the intercept, u_{0i} the random effect, and ϵ_{ij} is the residual error terms.

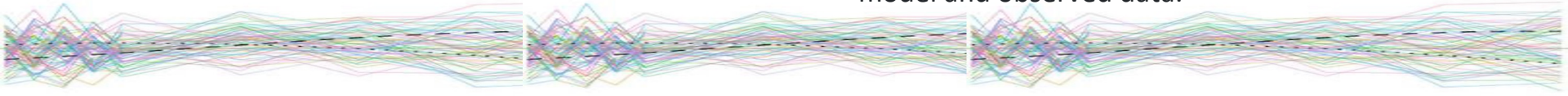
The model has the following distributional assumptions:

- $u_{0i} \sim N(0, \sigma^2_{u0})$
- $\epsilon_{ij} \sim N(0, \sigma^2)$
- $u_{0i} \perp \epsilon_{ij}$

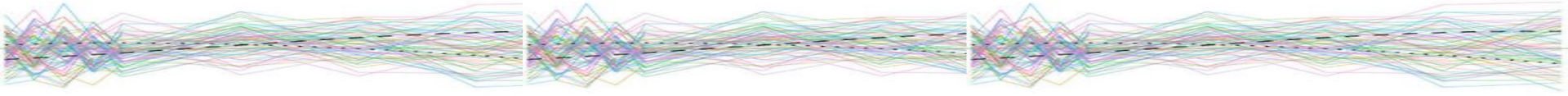
Interpretation

β_0 is the fixed effect representing the overall mean of the response variable Y_{ij}

- u_{0i} is a random variable of which we estimate the variance. It is the individual-specific deviation from β_0 that can be predicted from the model and observed data.

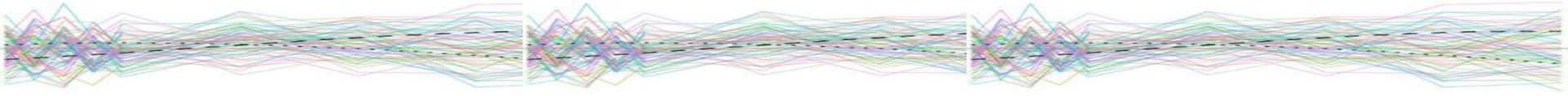


- The *conditional mean response* for a specific individual is $E[Y_{ij} | u_{0i}] = \beta_0 + u_{0i}$
- The *marginal mean response* in the population (i.e., averaged over all individuals in population) is $E[Y_{ij}] = \beta_0$



Estimation

- β coefficients can be estimated using *Maximum Likelihood* (ML) estimation
- ML does not take into account the degrees of freedom used for estimating fixed effects when estimating variance components
- This can lead to bias estimates (underestimation) for the variance components (small number of individuals).
Restricted Maximum Likelihood (REML) can overcome this limitation



Literature

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- Skrondal, A. & Rabe-Hesketh, S. (2004). *Generalized Latent Variable Modeling: Multilevel, longitudinal, and structural equation models*. Chapman & Hall/CRC Press.

