

Linear mixed effects models and where to find them

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Naming

Linear mixed effects models

Hierarchical linear models

Nested data models

Multilevel models

Random coefficient

Split-plot designs

Random-effects models

Linear mixed effects models

- Generalization of linear models (linear regression)
- Several IVs (predictors) —> Single DV (response)
- Multilevel IVs (nested data, between or within subjects)
- Fixed effects x random effects

Linear regression

- Y (response) \leftarrow X_i (predictors)

The diagram shows the linear regression equation $Y_i = \beta_0 + \beta_1 X_i + \epsilon_i$. Arrows point from labels to the corresponding parts of the equation: 'Dependent Variable' points to Y_i , 'Independent Variable' points to X_i , and 'Random Error term' points to ϵ_i . Two arrows from the text 'Fixed effect parameters' below point to β_0 and β_1 .

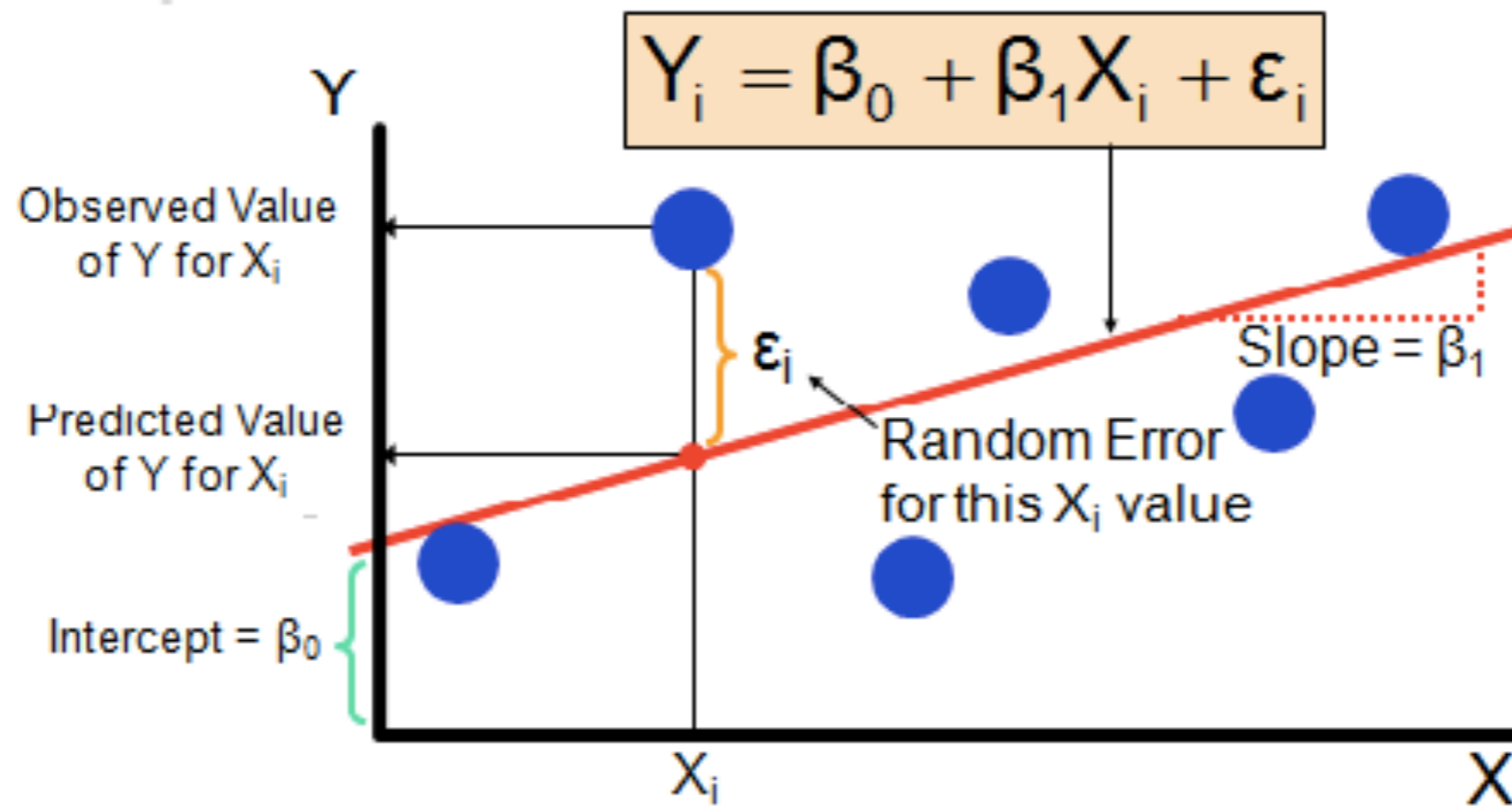
“Fixed effect parameters”

$$\text{observed data} \rightarrow y = b_0 + b_1 x_1 + b_2 x_2 + \dots + b_p x_p + \epsilon$$

$$\text{predicted data} \rightarrow y' = b_0 + b_1 x_1 + b_2 x_2 + \dots + b_p x_p$$

$$\text{error} \rightarrow \epsilon = y - y'$$

Linear regression



Random effect models

- Y (response) $\leftarrow X_i$ (predictors)
- Nested (hierarchical) data
- Constant random component

$$y_{ij} = b_0 + b_1 x_{ij} + v_i + e_{ij}$$

“Random effect”



Example:

$$Y_{ij} = \mu + \beta_1 \text{Sex}_{ij} + \beta_2 \text{Race}_{ij} + \beta_3 \text{ParentsEduc}_{ij} + U_i + W_{ij},$$

LME: Fixed + random effects

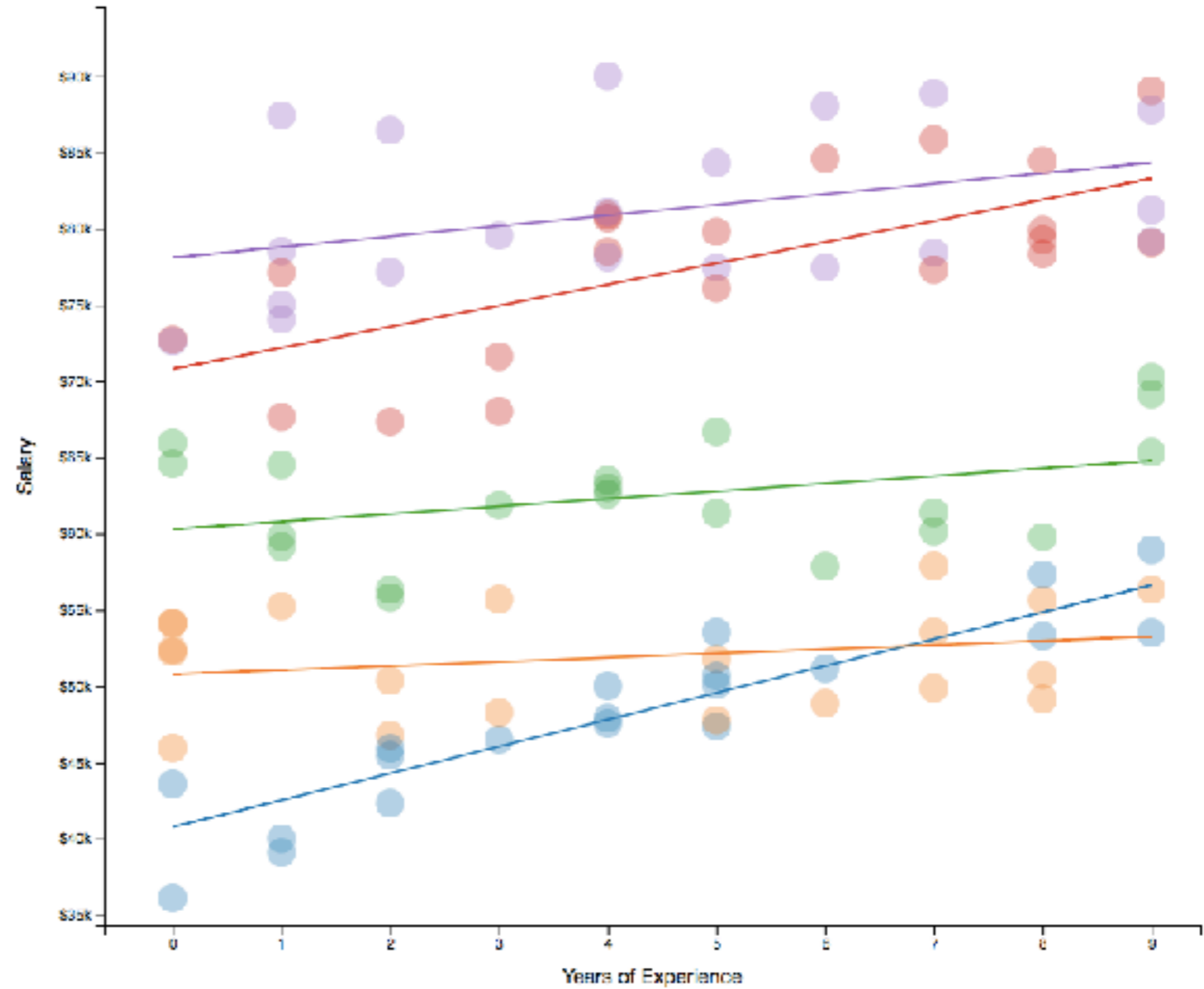
$$y_{ij} = b_0 + b_1 x_{ij} + v_i + e_{ij}$$

for $i \in \{1, \dots, n\}$ and $j \in \{1, \dots, m_i\}$ where

- $y_{ij} \in \mathbb{R}$ is the **response** for j -th measurement of i -th subject
- $b_0 \in \mathbb{R}$ is the **fixed intercept** for the regression model
- $b_1 \in \mathbb{R}$ is the **fixed slope** for the regression model
- $x_{ij} \in \mathbb{R}$ is the **predictor** for j -th measurement of i -th subject
- $v_i \stackrel{\text{iid}}{\sim} \mathcal{N}(0, \sigma_v^2)$ is the **random intercept** for the i -th subject
- $e_{ij} \stackrel{\text{iid}}{\sim} \mathcal{N}(0, \sigma_e^2)$ is a Gaussian **error term**

Nice example

<http://mfviz.com/hierarchical-models/>



How to use in R

- Packages *lme4* (Linear Mixed Effects), *nlme* (Non-Linear Mixed Effects), *asreml* (average spatial reml)
- Function *lmer()*

Score ~ Age + ParentsEduc + (1|Subject) + (1| School) + ε

```
>> lmer(Score ~ Age + ParentsEduc + (1|Subject) + (1| School),  
       data = datatable)
```

IOR data



Figure 1. Trial with the valid location of the target

IOR data

```
model1 <- lmer( MRT ~ probeOnset+ (1|NewSubject), dt)
```

```
model2 <- lmer( MRT ~ probeOnset+ validy+(1|NewSubject), dt)
```

```
model3 <- lmer( MRT ~ probeOnset+ validy+rFrame+(1|NewSubject), dt)
```

```
model3a <- lmer( MRT ~ early+ validy+early:validy:rFrame+(1|NewSubject), dt)
```

```
resultfull <- lmer( MRT ~ probeOnset*validy*rFrame + (1|NewSubject), dt)
```

Usage

- Nested data
- Repeated measures
- Longitudinal studies

LME vs ANOVA?

- Incorrect question! (ans: both)
- ANOVA can use LME
- *aov()*, *anova()*, *ezANOVA()*
aov(Y ~ B + Error(A/B), data=datatable)
- Or use *anova.lme()* from the *nlme* package

Resources

- Nice visualization <http://mfviz.com/hierarchical-models/>
- R function *lmer* good example <http://lme4.r-forge.r-project.org/book/Ch1.pdf>
- R *lmer* nice tutorial http://www.bodowinter.com/tutorial/bw_LME_tutorial2.pdf
- Math staff <http://www2.stat.duke.edu/~sayan/Sta613/2017/lec/LMM.pdf>