

Lab 4 - ARMD Trial

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For patient i ($i = 1, \dots, 234$) at time t ($t = 4, 12, 24, 52$ weeks)

- **Lab2 notation:** \rightarrow we add a random intercept b_{0i}

$$\text{VISUAL}_{it} = \beta_0 + \beta_1 \cdot \text{VISUAL0}_i + \beta_2 \cdot \text{TIME}_{it} + \quad (1)$$

$$+ \beta_3 \cdot \text{TREAT}_i + \beta_4 \cdot \text{TREAT}_i \cdot \text{TIME}_{it} + \quad (2)$$

$$+ b_{0i} + \epsilon_{it}, \quad (3)$$

$$\epsilon_{it} \sim \mathcal{N}(0, \sigma_{it}^2), \quad (4)$$

$$b_{0i} \sim \mathcal{N}(0, \sigma^2 d_{11}) \quad (5)$$

- **Lab3 and Lab4 notation:** \rightarrow with a random intercept b_{0i}

$$\underline{\text{VISUAL}}_i = \underline{\mathbb{X}}_i \underline{\beta} + \mathbb{1}_i b_{0i} + \underline{\epsilon}_i \quad (6)$$

$$\underline{\epsilon}_i \sim \mathcal{N}(\underline{0}, \sigma^2 \underline{\mathcal{R}}_i) \quad \text{where} \quad \underline{\mathcal{R}}_i = \underline{\Lambda}_i \underline{\mathcal{C}}_i \underline{\Lambda}_i \quad (7)$$

$$b_{0i} \sim \mathcal{N}(0, \sigma^2 d_{11}) \quad (8)$$

More in general: \rightarrow with random intercept and slopes $\underline{b}_i = [b_{0i} \ b_{1i} \ \dots]'$

$$\underline{\text{VISUAL}}_i = \underline{\mathbb{X}}_i \underline{\beta} + \underline{\mathbb{Z}}_i \underline{b}_i + \underline{\epsilon}_i \quad (9)$$

$$\underline{\epsilon}_i \sim \mathcal{N}(\underline{0}, \sigma^2 \underline{\mathcal{R}}_i) \quad \text{where} \quad \underline{\mathcal{R}}_i = \underline{\Lambda}_i \underline{\mathcal{C}}_i \underline{\Lambda}_i \quad (10)$$

$$\underline{b}_i \sim \mathcal{N}(\underline{0}, \sigma^2 \underline{\mathcal{D}}) \quad (11)$$

we know that $\underline{\mathcal{V}}_i = \underline{\mathbb{Z}}_i \underline{\mathcal{D}} \underline{\mathbb{Z}}_i' + \underline{\mathcal{R}}_i$

- with `'getVarCov(model, type = 'conditional')` we extract $\sigma^2 \underline{\mathcal{R}}_i$;
- with `'getVarCov(model, type = 'marginal')` we extract $\sigma^2 \underline{\mathcal{V}}_i$;
- with `'VarCorr(model)'` we extract $\sigma^2 \underline{\mathcal{D}}$ (also from the summary).

```
library(nlme)
lm2.form <- visual ~ visual0 + time + treat.f + treat.f:time
```

Homoscedastic residuals

Model 16.1 - Random intercept

$$\mathcal{D} = [d_{11}]$$

$$\mathcal{R}_i = \Lambda_i \mathcal{C}_i \Lambda_i = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{v}_i = \mathbb{Z}_i \mathcal{D} \mathbb{Z}_i' + \mathcal{R}_i = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} [d_{11}] \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\Rightarrow \mathbf{v}_i = \begin{bmatrix} 1 + d_{11} & d_{11} & d_{11} & d_{11} \\ d_{11} & 1 + d_{11} & d_{11} & d_{11} \\ d_{11} & d_{11} & 1 + d_{11} & d_{11} \\ d_{11} & d_{11} & d_{11} & 1 + d_{11} \end{bmatrix}$$

Note that the **implied marginal variance-covariance structure** is that of *compound symmetry* with a common correlation equal to $\rho = d_{11}/(1 + d_{11}) > 0$ since $d_{11} > 0$.

$$Var(\text{VISUAL}_{it}) = \sigma^2(d_{11} + 1)$$

```
fm16.1 <- lme(lm2.form, random = ~1|subject, data = armd)
```

Model 16.2 - Random intercept + slope

Model 16.2A - General D

$$\mathcal{D} = \begin{bmatrix} d_{11} & d_{12} \\ d_{21} & d_{22} \end{bmatrix}$$

$$\mathcal{R}_i = \Lambda_i \mathcal{C}_i \Lambda_i = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{v}_i = \mathbb{Z}_i \mathcal{D} \mathbb{Z}_i' + \mathcal{R}_i = \begin{bmatrix} 1 & 4 \\ 1 & 12 \\ 1 & 24 \\ 1 & 52 \end{bmatrix} \begin{bmatrix} d_{11} & d_{12} \\ d_{21} & d_{22} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 4 & 12 & 24 & 52 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$Var(\text{VISUAL}_{it}) = \sigma^2(d_{11} + 2d_{12}\text{TIME}_{it} + d_{22}\text{TIME}_{it}^2 + 1)$$

```
fm16.2A <- lme(lm2.form, random = ~1 + time | subject, data = armd)
```

Model 16.2B - Diagonal D

$$\mathcal{D} = \begin{bmatrix} d_{11} & 0 \\ 0 & d_{22} \end{bmatrix}$$

$$\mathcal{R}_i = \mathbf{\Lambda}_i \mathcal{C}_i \mathbf{\Lambda}_i = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{v}_i = \mathbb{Z}_i \mathcal{D} \mathbb{Z}_i' + \mathcal{R}_i = \begin{bmatrix} 1 & 4 \\ 1 & 12 \\ 1 & 24 \\ 1 & 52 \end{bmatrix} \begin{bmatrix} d_{11} & 0 \\ 0 & d_{22} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 4 & 12 & 24 & 52 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$Var(\text{VISUAL}_{it}) = \sigma^2(d_{11} + d_{22} \text{TIME}_{it}^2 + 1)$$

```
fm16.2B <- lme(lm2.form, random = list(subject = pdDiag(~time)), data = armd)
```

Heteroscedastic residuals: varPower()

Model 16.3 - Random intercept

$$\mathcal{D} = [d_{11}]$$

$$\mathcal{R}_i = \mathbf{\Lambda}_i \mathcal{C}_i \mathbf{\Lambda}_i = \begin{bmatrix} |\text{TIME}_{i1}|^{2\delta} & 0 & 0 & 0 \\ 0 & |\text{TIME}_{i2}|^{2\delta} & 0 & 0 \\ 0 & 0 & |\text{TIME}_{i3}|^{2\delta} & 0 \\ 0 & 0 & 0 & |\text{TIME}_{i4}|^{2\delta} \end{bmatrix}$$

$$\mathbf{v}_i = \mathbb{Z}_i \mathcal{D} \mathbb{Z}_i' + \mathcal{R}_i = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} [d_{11}] \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix} + \begin{bmatrix} |\text{TIME}_{i1}|^{2\delta} & 0 & 0 & 0 \\ 0 & |\text{TIME}_{i2}|^{2\delta} & 0 & 0 \\ 0 & 0 & |\text{TIME}_{i3}|^{2\delta} & 0 \\ 0 & 0 & 0 & |\text{TIME}_{i4}|^{2\delta} \end{bmatrix}$$

$$\Rightarrow \mathbf{v}_i = \begin{bmatrix} |\text{TIME}_{i1}|^{2\delta} + d_{11} & d_{11} & d_{11} & d_{11} \\ d_{11} & |\text{TIME}_{i2}|^{2\delta} + d_{11} & d_{11} & d_{11} \\ d_{11} & d_{11} & |\text{TIME}_{i3}|^{2\delta} + d_{11} & d_{11} \\ d_{11} & d_{11} & d_{11} & |\text{TIME}_{i4}|^{2\delta} + d_{11} \end{bmatrix}$$

$$Var(VISUAL_{it}) = \sigma^2(d_{11} + |TIME_{it}|^{2\delta})$$

```
fm16.3 <- update(fm16.1,
  weights = varPower(form = ~ time),
  data = armd)
```

Model 16.4 - Random intercept + slope

Model 16.4A - General D

$$\mathcal{D} = \begin{bmatrix} d_{11} & d_{12} \\ d_{21} & d_{22} \end{bmatrix}$$

$$\mathcal{R}_i = \mathbf{\Lambda}_i \mathcal{C}_i \mathbf{\Lambda}_i = \begin{bmatrix} |TIME_{i1}|^{2\delta} & 0 & 0 & 0 \\ 0 & |TIME_{i2}|^{2\delta} & 0 & 0 \\ 0 & 0 & |TIME_{i3}|^{2\delta} & 0 \\ 0 & 0 & 0 & |TIME_{i4}|^{2\delta} \end{bmatrix}$$

$$\mathcal{V}_i = \mathbb{Z}_i \mathcal{D} \mathbb{Z}_i' + \mathcal{R}_i = \begin{bmatrix} 1 & 4 \\ 1 & 12 \\ 1 & 24 \\ 1 & 52 \end{bmatrix} \begin{bmatrix} d_{11} & d_{12} \\ d_{21} & d_{22} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 4 & 12 & 24 & 52 \end{bmatrix} + \mathcal{R}_i$$

$$Var(VISUAL_{it}) = \sigma^2(d_{11} + 2d_{12}TIME_{it} + d_{22}TIME_{it}^2 + |TIME_{it}|^{2\delta})$$

```
fm16.4A <- update(fm16.3,
  random = ~1 + time | subject,
  data = armd)
```

Model 16.4B - Diagonal D

$$\mathcal{D} = \begin{bmatrix} d_{11} & 0 \\ 0 & d_{22} \end{bmatrix}$$

$$\mathcal{R}_i = \mathbf{\Lambda}_i \mathcal{C}_i \mathbf{\Lambda}_i = \begin{bmatrix} |TIME_{i1}|^{2\delta} & 0 & 0 & 0 \\ 0 & |TIME_{i2}|^{2\delta} & 0 & 0 \\ 0 & 0 & |TIME_{i3}|^{2\delta} & 0 \\ 0 & 0 & 0 & |TIME_{i4}|^{2\delta} \end{bmatrix}$$

$$\boldsymbol{\mathcal{V}}_i = \mathbb{Z}_i \boldsymbol{\mathcal{D}} \mathbb{Z}_i' + \boldsymbol{\mathcal{R}}_i = \begin{bmatrix} 1 & 4 \\ 1 & 12 \\ 1 & 24 \\ 1 & 52 \end{bmatrix} \begin{bmatrix} d_{11} & 0 \\ 0 & d_{22} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 4 & 12 & 24 & 52 \end{bmatrix} + \boldsymbol{\mathcal{R}}_i$$

$$Var(\text{VISUAL}_{it}) = \sigma^2(d_{11} + d_{22}\text{TIME}_{it}^2 + |\text{TIME}_{it}|^{2\delta})$$

```
fm16.4B <- update(fm16.3,
  random = list(subject = pdDiag(~time)), # Diagonal D
  data = armd)
```