

# Intro to mixed effects models

# Warm-up: class activity

[https://sta712-f22.github.io/class\\_activities/ca\\_lecture\\_39.html](https://sta712-f22.github.io/class_activities/ca_lecture_39.html)

- As  $\sigma_u^2$  increases
  - variance of the mixed effects does better relative to the other models
  - may see bias in  $\hat{\beta}$  if the model doesn't include groups
- Using fixed effects for each can result in increased variance of  $\hat{\beta}$ , particularly if  $X$  is correlated with group label

# Fitting mixed effects models

Mixed effects model:  
( $u \perp \varepsilon$ )

$$\begin{aligned} Y &= X\beta + Zu + \varepsilon \\ u &\sim N(0, G) \quad (\text{e.g., } G = \sigma_u^2 I) \\ \varepsilon &\sim N(0, R) \quad (\text{e.g., } R = \sigma_\varepsilon^2 I) \end{aligned}$$

$$\Rightarrow Zu + \varepsilon \sim N(0, ZGZ^T + R)$$

Marginal model:

$$\begin{aligned} Y &= X\beta + \varepsilon^* \\ \varepsilon^* &\sim N(0, \underbrace{ZGZ^T + R}_V) \end{aligned}$$

Assume  $V$  is known: use weighted least squares

$$\hat{\beta} = (X^T V^{-1} X)^{-1} X^T V^{-1} Y \quad (\text{inverse variance weighting})$$

If  $V$  is unknown: estimation is more complicated  
(jointly estimate  $\beta$  &  $V$ )

Case-control example:

$$Y_{ij} = \beta_0 + u_i + \beta_1 X_i + \varepsilon_{ij}$$

$$u_i \stackrel{iid}{\sim} N(0, \sigma_u^2)$$

$$\varepsilon_{ij} \stackrel{iid}{\sim} N(0, \sigma_\varepsilon^2)$$

Controls

Cases

$$X_i \in \{0, 1\}$$

control  $\uparrow$  case (treatment)

Groups

1, ..., m<sub>0</sub>

Groups

m<sub>0</sub>+1, ..., m

$$\Rightarrow X_1, \dots, X_{m_0} = 0$$

$$X_{m_0+1}, \dots, X_m = 1$$

Let  $n_i$  = # obs in group i

$$\text{Let } \bar{Y}_i \equiv \frac{1}{n_i} \sum_{j=1}^{n_i} Y_{ij}$$

Then  $\hat{\beta}_1$  (mixed effects model) =

$$\frac{n_i \bar{Y}_i}{n_i \sigma_u^2 + \sigma_\varepsilon^2} = \frac{\bar{Y}_i}{\sigma_u^2 + \frac{\sigma_\varepsilon^2}{n_i}}$$

$$\frac{\sum_{i=m_0+1}^m \frac{n_i \bar{Y}_i}{n_i \sigma_u^2 + \sigma_\varepsilon^2}}{\sum_{i=m_0+1}^m \frac{n_i}{n_i \sigma_u^2 + \sigma_\varepsilon^2}}$$

$$- \frac{\sum_{i=1}^{m_0} \frac{n_i \bar{Y}_i}{n_i \sigma_u^2 + \sigma_\varepsilon^2}}{\sum_{i=1}^{m_0} \frac{n_i}{n_i \sigma_u^2 + \sigma_\varepsilon^2}}$$

If instead we fit a linear model w/out

$\hat{\beta}_1$

$$= \frac{\sum_{i=m_0+1}^m n_i \bar{Y}_i}{\sum_{i=m_0+1}^m n_i}$$

group effect:

$$\frac{\sum_{i=1}^{m_0} n_i \bar{Y}_i}{\sum_{i=1}^{m_0} n_i}$$

← This gives potentially much more weight to groups with more observations

# Assumptions

$$Anxiety_{ij} = \beta_0 + u_i + \beta_1 SmallEnsemble_{ij} + \beta_2 Solo_{ij} + \varepsilon_{ij}$$

$$u_i \stackrel{iid}{\sim} N(0, \sigma_u^2) \quad \varepsilon_{ij} \stackrel{iid}{\sim} N(0, \sigma_\varepsilon^2)$$

What assumptions does this mixed effects model make?

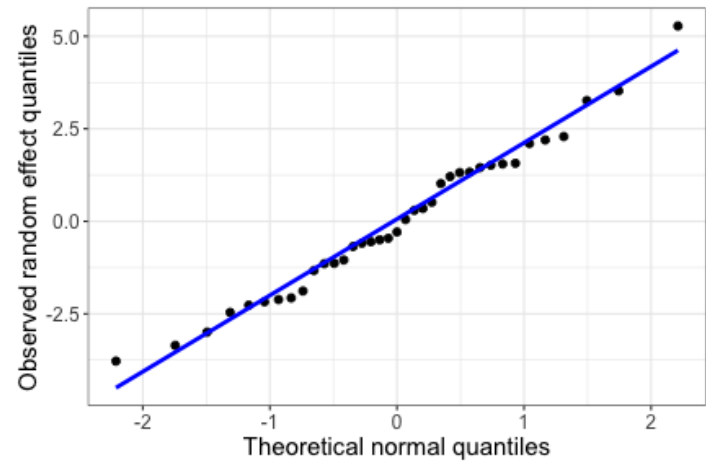
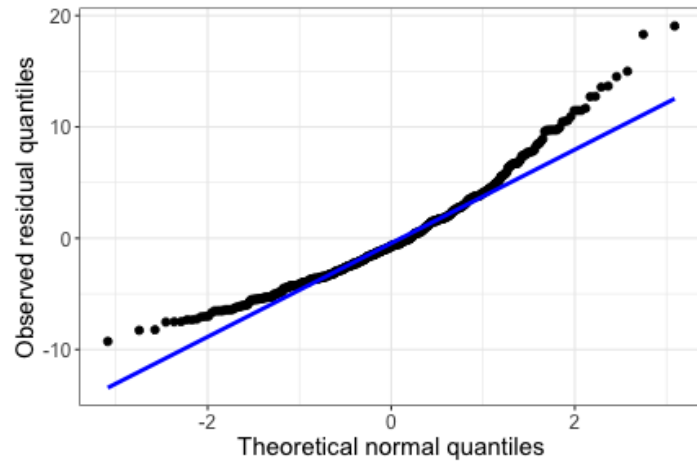
## Assessing normality

$$Anxiety_{ij} = \beta_0 + u_i + \beta_1 SmallEnsemble_{ij} + \beta_2 Solo_{ij} + \varepsilon_{ij}$$

$$u_i \stackrel{iid}{\sim} N(0, \sigma_u^2) \quad \varepsilon_{ij} \stackrel{iid}{\sim} N(0, \sigma_\varepsilon^2)$$

How should we check the normality assumption?

# QQ plots



## Changing the model

$$Anxiety_{ij} = \beta_0 + u_i + \beta_1 SmallEnsemble_{ij} + \beta_2 Solo_{ij} + \varepsilon_{ij}$$

$$u_i \stackrel{iid}{\sim} N(0, \sigma_u^2) \quad \varepsilon_{ij} \stackrel{iid}{\sim} N(0, \sigma_\varepsilon^2)$$

How could we change the model to allow the effect of performance type to differ between musicians?

$$Anxiety_{ij} = (\beta_0 + u_i) + (\beta_1 + v_i) Small_{ij} + (\beta_2 + w_i) Solo_{ij}$$

$$\varepsilon_{ij} \stackrel{iid}{\sim} N(0, \sigma_\varepsilon^2)$$

$$\begin{bmatrix} u_i \\ v_i \\ w_i \end{bmatrix} \stackrel{iid}{\sim} N \left( \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \right.$$

$$\left. \begin{bmatrix} \sigma_u^2 & \rho_{uv} \sigma_u \sigma_v & \rho_{uw} \sigma_u \sigma_w \\ & \sigma_v^2 & \rho_{vw} \sigma_v \sigma_w \\ & & \sigma_w^2 \end{bmatrix} \right) + \varepsilon_{ij}$$



# Fitting the model

```
m2 <- lmer(na ~ perform_type + (perform_type|id),
           data = music)
summary(m2)
```

...

## Random effects:

##	Groups	Name	$\hat{\sigma}_u^2$	Variance	Std.Dev.	Corr
##	id	(Intercept)		3.986	1.997	
##		perform_typeSmall Ensemble		2.019	1.421	-0.43
##		perform_typeSolo		1.017	1.008	0.74
##	Residual			21.288	4.614	

## Number of obs: (497), groups: (id, 37)

## Fixed effects:

##		Estimate	Std. Error	t value
##	(Intercept)	15.0503	0.5436	27.685
##	perform_typeSmall Ensemble	0.6996	0.7410	0.944
##	perform_typeSolo	2.0134	0.5671	3.550

...

# Prediction

What is the estimated anxiety for Musician 1 before a solo performance?

```
coef(m2)
```

```
...  
## $id  
##      (Intercept) perform_typeSmall Ensemble perform_typeSolo  
## 1      12.37560          0.84623321          0.6590148  
## 2      13.61693          0.30915635          1.0413577  
## 3      12.86707          1.31366273          1.1674007  
...
```

$$\underbrace{12.38}_{\hat{\beta}_0 + \hat{u}_1} + \underbrace{0.66}_{\hat{\beta}_2 + \hat{w}_1}$$