

# Correlated / Clustered / Longitudinal Data -

4/14/22

n "clusters"

$i = 1, 2, \dots, n$

$t_i$  = index of  $t^{\text{th}}$  measurement  
on cluster  $i$

$$\underline{Y}_i = \begin{bmatrix} Y_{i1} \\ Y_{i2} \\ \vdots \\ Y_{it_i} \end{bmatrix}$$

$$\text{Cov}(\underline{Y}_i) = \begin{bmatrix} \text{Var}(Y_{i1}) & \text{Cov}(Y_{i1}, Y_{i2}) & \dots \\ \text{Cov}(Y_{i2}, Y_{i1}) & \text{Var}(Y_{i2}) & \dots \\ \vdots & \vdots & \ddots \\ \text{Cov}(Y_{it_i}, Y_{i1}) & \text{Cov}(Y_{it_i}, Y_{i2}) & \text{Var}(Y_{it_i}) \end{bmatrix}$$

BEE  $\rightarrow$  Model 1<sup>st</sup> two moments -

$$E(\underline{Y}_i) \quad \text{and} \quad \text{Cov}(\underline{Y}_i)$$

- no distributional assumption

↑  
primary interest

↑ nuisance  
parameters

$\rightarrow$  Robust to choice of  
covariance structure

Common Correlation Structures:  $\text{Corr}(\underline{Y}_i) = \begin{bmatrix} 1 & \rho_{12} & \rho_{13} & \dots \\ \rho_{21} & 1 & \dots & \dots \\ \vdots & \vdots & \ddots & \dots \\ \rho_{23} & \dots & \dots & 1 \end{bmatrix}$

Independence:  $\text{Corr}(\underline{Y}_i) = I_{t_i}$   
 $= \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & \vdots \\ \vdots & \vdots & \ddots & 1 \\ 0 & \dots & \dots & 1 \end{bmatrix} \quad t_i = 3$

Exchangeable / Compound Symmetric:  $\text{Corr}(\underline{Y}_i) = \begin{bmatrix} 1 & \rho & \rho \\ \rho & 1 & \rho \\ \rho & \rho & 1 \end{bmatrix}$

AR(1) - autoregressive:

$$\text{Corr}(\underline{Y}_i) = \begin{bmatrix} 1 & \rho & \rho^2 \\ \rho & 1 & \rho \\ \rho^2 & \rho & 1 \end{bmatrix}$$

Unstructured:  $\text{Corr}(\underline{Y}_i) = \begin{bmatrix} 1 & \rho_{12} & \rho_{13} \\ \rho_{12} & 1 & \rho_{23} \\ \rho_{13} & \rho_{23} & 1 \end{bmatrix}$

## Option 2: Subject-specific models

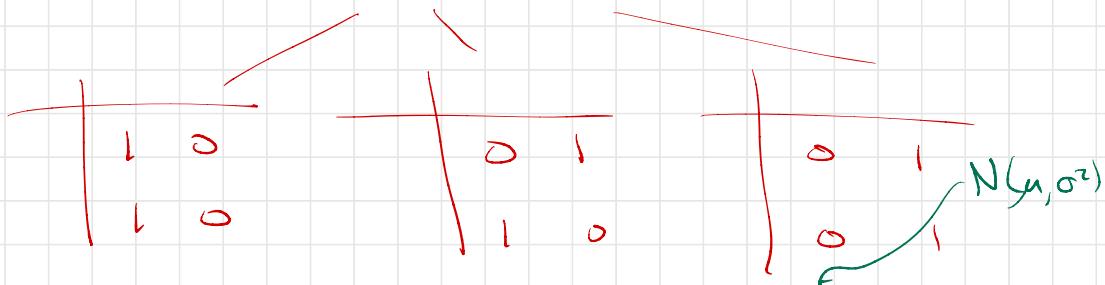
- add random/fixed effect per cluster

Partial table for each subject:

Example

		Asthma	
		Yes	No
Age	13	1	0
	20	0	1

→ total # observations  
→ # repeated measures



Model:  $\text{logit}(\text{P}(Y_{it} = 1)) = \alpha_i + \beta X_{it}$

person  $i$  at time  $t$

$$\mu + \tau_i \sim N(0, \sigma^2)$$

Compare to:  
GEE

$$\text{logit}(\text{P}(Y_t = 1)) = \alpha + \delta X_t \quad t=1, 2$$

Correlation within subject is induced

## Example - Asthma

GEE:  $Y_t = \begin{cases} 1 & \text{asthma at time } t \\ 0 & \text{else} \end{cases}$

Assumptions:  $\sim E(Y_t)$

$$\text{logit}(P(Y_t=1)) = \alpha + \beta X_t$$

$$X_t = \begin{cases} 1 & \text{age 20} \\ 0 & \text{age 13} \end{cases}$$

$$\text{Cov}\left(\begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix}\right) = \begin{bmatrix} \sigma^2 & \tau \\ \tau & \sigma^2 \end{bmatrix}$$

GLMM:  $Y_{it} = \begin{cases} 1 & \text{asthma at time } t \text{ for person } i \\ 0 & \text{else} \end{cases}$

$$\text{logit}(P(Y_{it}=1))$$

$$= \mu + \alpha_i + \beta X_{it}$$

$$\alpha_i \stackrel{\text{iid}}{\sim} N(0, \sigma^2)$$

$$Y_{it} | \alpha_i \sim \text{Bin}(1, E(Y_{it} | \alpha_i))$$

$$\text{logit}(E(Y_{it} | \alpha_i)) = \mu + \alpha_i + \beta X_{it}$$

Aside: (deriving link):

$$E(Y_{it} | \alpha_i) = \mu + \alpha_i + \beta X_{it}$$

$$\Rightarrow E(Y_{ic}) = E[E(Y_{it} | \alpha_i)] = \mu + E(\alpha_i) + \beta X_{it} \\ = \mu + \beta X_{it}$$

Issue:  $g(E(Y_{it} | \alpha_i)) = \mu + \alpha_i + \beta X_{it}$  ← true

$$g(E(Y_{ic})) \neq E[g(E(Y_{it} | \alpha_i))]$$

$$g(E(E(Y_{it} | \alpha_i)))$$

## Fitted Models

GEE:

$$\text{logit}(\hat{P}(Y_t=1)) = -1.782 - 0.248 X_t$$

GLMM:

$$\text{logit}(\hat{P}(Y_{it}=1 | \alpha_i)) = -8.087 + \hat{\alpha}_i - 1.032 X_{it}$$

Odds Ratio:  $\left( \frac{P(\text{asthma} | \text{age } 20)}{P(\text{no asthma} | \text{age } 20)} \right) / \left( \frac{P(\text{asthma} | \text{age } 13)}{P(\text{no asthma} | \text{age } 13)} \right)$

SEE: -0.248

$e^{-0.248} = 0.780$



Estimated 22% decrease  
in the odds of asthma  
at age 20 compared to age 13.

population-averaged

GLMM: -1.032

$e^{-1.032} = 0.356$



Estimated 64%  
decrease

subject-specific

Simple case → closed form solutions to those  
estimated odds ratios

GEE:

$$\hat{OR}_{age 13/20} = \frac{ad}{bc} = 1.282$$

$$\Rightarrow \hat{OR}_{age 20/13} = \frac{1}{1.282} = 0.780$$

GLMM:  $\hat{OR} = n_{12}/n_{21}$

$$= 22/8 = 2.75$$

$$\Rightarrow \hat{OR} = \frac{1}{2.75} = 0.364$$

		Original data:	
		Age 20	Asthma
Age	No.	<del>Asthma</del>	
		$n_{11}=50$	$n_{12}=22$
13	No.	$n_{21}=8$	$n_{22}=420$

Create Marginal Table:

Age	No.	Asthma	
		Yes	No
13	No.	a=72	b=428
20	No.	c=58	d=442