Biostatistics 140.656, 2018-19 Lab 4 Solution (Partial)

Topics:

- Mixed logistic regression models for hierarchical data with more than 2 levels
- Interpretation of parameters from mixed logistic regression model with multiple random intercepts
- Interpretation of parameters from mixed logistic regression model with random slopes

Learning Objectives:

Students who successfully complete this lab will be able to:

- Explore data with 3 levels of nesting
- Interpret coefficients for a level 1 exposure in a mixed logistic model with random intercepts defined at level 2 and 3
- Interpret results from a mixed logistic model including multiple random intercept and random slope

Scientific Background:

You will be analyzing data from a survey conducted in Guatemala in 1987. The survey identified a nationally representative sample of 5160 mothers, between 15 and 44 years of age, with the primary purpose of understanding factors that could affect the immunization status of children who were born in the previous 5 years and alive at the time of the interview.

The data available represent 2,158 children (level 1) aged 1 – 4 years of age from 1595 mothers (level 2) from within 161 communities (level 3). The 2,158 children received at least one immunization. The outcome of interest is whether the child received the full set of immunizations.

Starting in 1986, the government of Guatemala undertook a series of campaigns to immunize the population against major childhood illnesses. The immunization campaign visited most of the country and often located children in their own households. The full set of immunizations at the time of the campaign included three doses of DPT vaccine (against diphtheria, whooping cough, and tetanus), three doses of polio vaccine, one dose of BCG (antituberculosis) and one dose of the measles vaccine.

The survey conducted in 1987 offers an opportunity to evaluate the likelihood of children receiving the full set of immunization during both absence (pre-1986) vs. presence (1986-7) of the campaign (i.e. a natural experiment). An important variable is whether the child was at least 2 years old at the time of the survey/interview, in which case the child was eligible to receive all immunizations during the campaign. If this variable is associated with immunization status, there is some indication that the campaign worked.

In your final exam, you will be conducting a series of analyses to address questions that are relevant to government health officials both from the prospective of improving the immunization coverage rate but also to explore factors related to differences in the odds of full immunization that could provide insight into how to improve the design and implementation of future campaigns.

In this lab, you will:

- 1. Summarize the three-level nested structure of the data
- 2. Fit and interpret a random intercept only logistic regression model within this three-level nested data example
- 3. Estimate the effect of the immunization campaign within a random effects logistic regression model
- 4. Estimate the heterogeneity of the effect of the immunization campaign across the Guatemalan communities and determine if community characteristics explain observed heterogeneity.

Data:

The data set is called **guatemala.csv**, which can be downloaded directly from our website. The dataset contains children i nested in mother j nested in community k. It contains the following subset of variables.

Level 1 (children)

- immun: dummy variable for child being immunized, the response variable.
- kid2p: child at least 2 years old at the time of the interview (indicator for exposure to the campaign or not)

Level 2 (mother)

- mom: identifier for mother
- Ind: Indigenous Ethnicity (indigenous vs. not)

Level 3 (community)

- cluster: identifier for communities
- rural: dummy variable for community being rural
- pcInd: percent of indigenous mothers in the community (ranges from 0 to 1)

Brief EDA of the hierarchical clustering:

How many communities are in the study, how many mothers and how many children?

```
. codebook cluster mom kid
cluster
                                                                                       (unlabeled)
                 type: numeric (int)
                range: [1,240]
                                                     units: 1
        unique values: 161
                                              missing .: 0/2,158
             mean: 145.858
std. dev: 59.3406
          percentiles: 10% 25% 50% 75% 63 94 148 202
                 type: numeric (int)
                range: [2,2782] units: 1
values: 1,595 missing .: 0/2,158
        unique values: 1,595
                 mean:
                         1502.63
             mean: 1502.63
std. dev: 751.435
          std. dev: 751.435

percentiles: 10% 25% 50% 75% 498 859 1471.5 2208
                                                                                        (unlabeled)
                 type: numeric (int)
                range: [2,4627] units: 1
values: 2,158 missing .: 0/2,158
         unique values: 2,158
             mean: 2445.98
std. dev: 1267.71
          percentiles: 10%
                                                                      90%
```

We have 161 communities; 1595 mothers and 2158 children.

How many mothers in each community?

The average number of mothers per community was roughly 10 with a range from 1 to 37.

How many children in each community?

- . bys cluster: egen cluster_kids = count(cluster)
- . summ cluster_kids if cluster_counter==1

Variable	Obs	Mean	Std. Dev.	Min	Max
cluster_kids	 161	13.40373	8.750271	 1	55

The number of children per community ranges from 1 to 55. Average number of children per community is 13.

How many children per mother?

. tab num_kids if kid_counter==1

num_kids	Freq.	Percent	Cum.
1 2 3	1,063 501 31	66.65 31.41 1.94	66.65 98.06 100.00
Total	1,595	100.00	

The number of children per mother ranges from 1 to 3 with 67% of the mothers contributing data from a single child.

Brief EDA of the primary outcome

What is the sample prevalence of immunization?

. summ immun

Variable	Obs	Mean	Std. Dev.	Min	Max
immun	2,158	.4467099	.4972673	0	1

What is the sample prevalence of immunization by study period (during or post campaign)?

Variable	0bs	Mean	Std. Dev.	Min	Max
immun	492	.2845528	.4516604	0	1
Variable	Obs	Mean	Std. Dev.	Min	Max
immun	1,666	.4945978	.5001209	0	1

NOTE: I will be providing you with all the necessary regression output relating to the mixed logistic regression models I want you to consider. Most of these regression models take a bit of time to run. To save you this time, I will provide the output and your goal is to focus on the interpretation.

1. Three-level random intercepts only model

To start we will ignore all the covariates in the model and simply decompose the variation in log odds of receiving the full course of immunizations into that attributable to differences across mothers within a given community and across communities.

Let Y_{ijk} be the indicator for completing the full course of immunizations for kid k from mom j within community i. Then the random intercept only model is given by:

$$log\left(\frac{Pr(y_{ijk}=1)}{1-Pr(y_{ijk}=1)}\right) = \beta_0 + b_i + b_{ij}, \ b_i \sim N(0,\sigma^2), b_{ij} \sim N(0,\tau^2), Cov(b_i,b_{ij}) = 0$$

In the model above, we allow each child to have his/her own log odds of receiving the full course of immunizations which depends on their mother and community membership.

The fit of this model is given below:

```
. meqrlogit immun || cluster: || mom: , intp(12)
Mixed-effects logistic regression
                                Number of obs =
                                                 2,158
| No. of Observations per Group Integration
Group Variable | Groups Minimum Average Maximum Points
cluster | 161 1 13.4 55 12
mom | 1,595 1 1.4 3 12
                                 Wald chi2(0) =
Log likelihood = -1396.5023
                                 Prob > chi2
    immun | Coef. Std. Err. z P>|z| [95% Conf. Interval]
    _cons | -.3733179 .1301502 -2.87 0.004 -.6284075 -.1182283
 Random-effects Parameters | Estimate Std. Err. [95% Conf. Interval]
cluster: Identity
           var(_cons) | 1.329794 .3547759
                                        .7883008 2.243244
mom: Identity
           var(_cons) | 4.14361 .9381624
                                       2.658623 6.458044
______
LR test vs. logistic model: chi2(2) = 174.06
                                       Prob > chi2 = 0.0000
```

a. Interpret the estimated value of β_0 and $\exp(\beta_0)/[1 + \exp(\beta_0)]$.

The estimated value of β_0 is -0.37; this represents the log odds of receiving the full course of immunizations for children of the average mom (i.e. b_{ij} =0) within the average community (i.e. $b_i = 0$).

The estimate value of $\exp(\beta_0)/[1 + \exp(\beta_0)]$ is approximately 41%; this represents the probability of receiving the full course of immunizations for children of the average mom within the average community.

b. Create an interval that contains the probability of a child receiving the full course of immunizations among children of 95% of the moms from the "average" community.

The "average" community is defined by setting $b_i = 0$.

Within any community *i*, the random intercept defined at the mom level allows for the log odds of receiving the full course of immunizations to vary from mom to mom.

Therefore, our calculation would be:

$$\beta_0 + /-1.96 \times \tau \rightarrow -0.37 + /-1.96 * sqrt(4.14) \rightarrow -4.36 \text{ to } 3.62$$

Converting this to an interval representing the probability of receiving the full course of immunizations for 95% of the moms from the "average" community yields 0.013 to 0.974.

c. Create an interval that contains the probability of a child receiving the full course of immunizations among children of 95% of the moms in Guatemala.

Here we are considering both sources of variation, across community and across moms within a community.

Therefore, our calculation would be

$$\beta_0 + / -1.96 \times \sqrt{\sigma^2 + \tau^2} \rightarrow -0.37 + / -1.96 * sqrt(4.14+1.31) \rightarrow -4.95 \text{ to } 4.21$$

Converting this to an interval representing the probability of receiving the full course of immunizations for 95% of the moms in Guatemala yields 0.007 to 0.985.

d. Intra-class correlation coefficients:

Applying the latent variable formulation of the mixed logistic model below implies that $Var(Y_{ijk}) = \sigma^2 + \tau^2 + \frac{\pi^2}{3}$.

This is derived by the following:

$$Y_{ijk} = 1 \rightarrow y^*_{ijk} = \beta_0 + b_i + b_{ij} + \varepsilon_{ijk} > 0,$$

 $b_i \sim N(0, \sigma^2), b_{ij} \sim N(0, \tau^2), \varepsilon_{ijk} \sim independent \ Logistic, Var(\varepsilon_{ijk}) = \frac{\pi^2}{3},$ and lastly b_i, b_{ij} and ε_{ijk} independent.

We may want to compute various intraclass correlation coefficients within this random intercept only model.

- What is the correlation between the binary indicator of receiving the full course of immunizations for two children from the same mother, i.e. $Corr(Y_{ijk}, Y_{ijm})$ (i.e. the intraclass correlation for moms)

HINT:
$$\frac{Cov(Y_{ijk}, Y_{ijm})}{\sigma^2 + \tau^2 + \frac{\pi^2}{3}} = \frac{Cov(b_i + b_{ij}, b_i + b_{ij})}{\sigma^2 + \tau^2 + \frac{\pi^2}{3}} = \frac{\sigma^2 + \tau^2}{\sigma^2 + \tau^2 + \frac{\pi^2}{3}} = \frac{4.14 + 1.33}{4.14 + 1.33 + 3.29} = 0.62$$

- What is the correlation between the binary indicator of receiving the full course of immunizations for two children from different mothers within the same community, i.e. $Corr(Y_{ijk}, Y_{imn})$ (i.e. the intraclass correlation for community)?

$$\frac{Cov(Y_{ijk}, Y_{imn})}{\sigma^2 + \tau^2 + \frac{\pi^2}{3}} = \frac{Cov(b_i + b_{ij}, b_i + b_{im})}{\sigma^2 + \tau^2 + \frac{\pi^2}{3}} = \frac{\sigma^2}{\sigma^2 + \tau^2 + \frac{\pi^2}{3}} = \frac{1.33}{4.14 + 1.33 + 3.29} = 0.15$$

2. Three-level random intercept model with level-1 covariate

Now consider adding the primary covariate (kid2p) to the model. Recall, this is the indicator for whether the child was exposed to the immunization campaign or not.

$$log\left(\frac{Pr(y_{ijk}=1)}{1-Pr(y_{ijk}=1)}\right) = \beta_0 + b_i + b_{ij} + \beta_1 kid2p_{ijk}, \ b_i \sim N(0,\sigma^2), b_{ij} \sim N(0,\tau^2), Cov(b_i,b_{ij}) = 0$$

In the model above, we allow each child to have his/her own log odds of receiving the full course of immunizations which depends on their mother and community membership; in addition, we assume that the effect of the immunization campaign is to change the log odds of receiving the full course of immunizations by the same factor (β_1) for each child.

The fit of this model is provided below:

. meqrlogit immun kid2p || cluster: || mom: , intp(12)

Mixed-effects logistic regression Number of obs = 2,158

| No. of Observations per Group Integration

Group Variable | Groups Minimum Average Maximum Points

cluster | 161 1 13.4 55 12

mo	om 	1,595	5 	1		1.4	3		12
Log likelihood	d = -	1353.5579					chi2(1) chi2		61.26 0.0000
immun	 	Coef.	Std.	Err.	z	P> z	[95%	Conf.	Interval]
		.668822 .722079							2.086731 -1.265041
Random-effe	cts P	arameters	 	Estimate	e Std	 . Err.	[95%	Conf.	Interval]
cluster: Ident	-	var(_cons)	+ 	1.592903	L .43	22593	.9358	3395	2.71129
mom: Identity		var(_cons)	+)	5.229659	9 1.1	83445	3.356	5224	8.148842
LR test vs. lo	 ogist	ic model:	chi2	(2) = 189	 9.90		Prob	> chi	2 = 0.0000
Note: LR test . megrlogit,		onservativ	ve an	d provide	ed only	for re	ference.		
Mixed-effects		stic regre	essio	n		Number	of obs	=	2,158
Group Variab	 le	No. of Groups	 E ≅	Observ Minimum	vations Aver	per Gr	oup Maximum	Inte	gration
cluste	+- er	No. of Groups 	s l	Minimum	Aver 1	age 3.4	oup Maximum 55 3	Integ	gration oints 12 12
cluste	+- er om	Groups 161 1,595	5 L 5 	Minimum 1	Aver 1	age 3.4 1.4 	Maximum	P(oints 12
cluste mo	er om 	Groups 16: 1,599 1353.5579	5 1 5 	Minimum1 1 1	Aver 1	3.4 1.4 Wald c	Maximum 55 3 chi2(1) chi2	P(oints 12 12 61.26
cluste mo Log likelihood immun kid2p _cons	er om d = - odd + 5	Groups 16: 1,59! 1353.5579 s Ratio 305913 1786943	Std.	Minimum 1 1 Err 1341 6691	Aver 1 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	Wald c Prob >	Maximum 55 3 hi2(1) chi2 [95% 3.493 .1131	Po	oints 12 12 61.26 0.0000 Interval] 8.058526 .2822276
cluste mo Log likelihood immun kid2p _cons Note: _cons es	er om d = - Odd + 5 . stima	Groups 16: 1,59! 1353.5579 s Ratio305913 1786943 tes basel:	Std. Std. 1.13	Minimum 1 1 Err 1341 6691 dds (cond	Aver 1 1 2 7.83 -7.38 ditiona	Wald c Prob > 0.000 0.000	Maximum 55 3 hi2(1) chi2 [95% 3.493 .1131 ro randon	Po	oints 12 12 61.26 0.0000 Interval] 8.058526 .2822276 cts).
cluste mo Log likelihood immun kid2p _cons Note: _cons es	er om d = - Odd + 5 . stima	Groups 16: 1,59! 1353.5579 s Ratio305913 1786943 tes basel: arameters	Std. Std. 1.13 .041	Minimum 1 1 Err 1341 6691 dds (cond	Aver 1 1 2 7.83 -7.38 ditiona	Wald comprobes the second seco	Maximum 55 3 chi2(1) chi2 [95% 3.493 .1131 ro random	Po	oints 12 12 61.26 0.0000 Interval] 8.058526 .2822276 cts) Interval]
cluste mo Log likelihood immun kid2p cons Note: cons es Random-effed cluster: Ident	er om d = - Odd + 5 stima	Groups 16: 1,59! 1353.5579 s Ratio 305913 1786943 tes basel: arameters var(_cons	Std Std 1.13 .041	Minimum 1 1 Err 1341 6691 dds (cond	Aver 1 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	Wald of Prob > 0.000 0.000 1 on zee 22593	Maximum 55 3 chi2(1) chi2 [95% 3.493 .1131 cro random	Po	oints 12 12 61.26 0.0000 Interval] 8.058526 .2822276 cts) Interval]
clusted model in the control of the	er om d = - Odd + 5 stima	Groups 16: 1,59! 1353.5579 s Ratio305913 1786943 tes basel: arameters var(_cons	Std Std 1.13 .041 ine o	Minimum 1 1 Err 1341 6691 dds (cond	Aver 1 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	Wald comprobes the second seco	Maximum 55 3 hi2(1) chi2 [95% 3.493 .1131 ro random [95%9358	Po	oints 12 12 61.26 0.0000 Interval] 8.058526 .2822276 cts) Interval]

a. Interpret the estimated value of $\exp(\beta_1)$.

For a given mom within a community in Guatemala, the odds of her completing the full course of immunizations for a child during the immunization campaign is 5.30 times the odds of completing the full course of immunizations for a child prior to the immunization campaign (95% CI: 3.49 to 8.06).

b. Describe an approach to estimate (ignore inference for now) the population level / marginal association between the odds of receiving the full course of immunizations and the immunization campaign.

We could fit a marginal model that would take the form:

$$log\left(\frac{\Pr(y_{ijk}=1)}{1-\Pr(y_{ijk}=1)}\right) = \alpha_0 + \alpha_1 kid2p_{ijk},$$

 $Corr(Y_{ijk}, Y_{ijm}) = \rho_2$, $Corr(Y_{ijk}, Y_{imn}) = \rho_1$, i.e. children within moms are exchangeable and moms within communities are exchangeable.

3. Three-level random intercepts plus random slope model

NOTE: No solution will be provided below as these questions or similar will be part of your final exam.

If you are a member of the health department in Guatemala, you may be interested in understanding if there was heterogeneity in the effect of the campaign across the communities.

Quantifying the heterogeneity may be of central interest as well as identifying communities where the campaign was more or less effective could lead to subsequent targeted changes in implementation of future campaigns.

Extend the model from part 2 to include a community level random slope for kid2p.

$$\log\left(\frac{\Pr(y_{ijk}=1)}{1-\Pr(y_{ijk}=1)}\right) = \beta_0 + b_{0i} + b_{0ij} + (\beta_1 + b_{1i})kid2p_{ijk},$$

$$\begin{split} &b_{0i} \sim N(0, \sigma_0^2), b_{1i} \sim N(0, \sigma_1^2), Cov(b_{0i}, b_{1i}) = \tau_{01}, \\ &b_{0ij} \sim N(0, \tau^2), \\ &Cov(b_{0i}, b_{0ij}) = 0, Cov(b_{1i}, b_{0ij}) = 0 \end{split}$$

The fit of this model is presented below:

meqrlogit immun kid2p || cluster: kid2p, cov(uns) || mom: , intp(12)

Mixed-effects logistic regression Number of obs = 2,158

Group Variable	No. of	Observ	ations per	Group	Integration
	Groups	Minimum	Average	Maximum	Points
cluster	161	1	13.4	55	12
mom	1,595	1	1.4	3	12

immun	!				[95% Conf.	Interval]
kid2p	1.933637	.2966054	6.52	0.000		

Random-effects Parameters | Estimate Std. Err. [95% Conf. Interval]

mom: Identity

va	r(_cons)	5.889154	1.3969	3.699551	9.37469
LR test vs. logistic	model: chi2	2(4) = 199.3	32	Prob > ch	i2 = 0.0000
. meqrlogit, or					
Mixed-effects logist	ic regressio	n	Number	of obs =	2,158
 Group Variable	No. of Groups	Observat	ions per Gro Average I	oup Int Maximum	egration Points
	 161	 1	13.4	 55	12
mom	1,595		1.4	3	12
Log likelihood = -13	48.8518			ni2(1) = chi2 =	
immun Odds	 Ratio Std.	Err.	z P> z	[95% Conf	. Interval]
kid2p 6.9 _cons .13	14616 2.05 71947 .042	60912 6 86523 -6	.52 0.000 .39 0.000	3.866314 .0745948	
Note: _cons estimate	s baseline c	odds (condit	cional on ze	ro random eff	ects).
Random-effects Par	ameters	Estimate	Std. Err.	[95% Conf	. Interval]
cluster: Unstructure	 d				
	r(kid2p)			.6896979	
	r(_cons)			1.682786	
COV(K1Q2)	p,_cons)	-1.981638 	1.14151	-4.218957 	.2556801
mom: Identity va	r(_cons)	5.889154	1.3969	3.699551	9.37469
LR test vs. logistic	model: chi2	2(4) = 199.3	32	Prob > ch	i2 = 0.0000
Note: LR test is con	servative an	nd provided	only for re	ference.	

a. Interpret the estimated value of $\exp(\beta_1)$.

b. Provide an interval that contains the effect of the campaign for 95% of the communities in Guatemala.

c. Estimate the $Corr(b_{0i},b_{1i})$ and interpret this value within the context of the problem