# Biostatistics 140.655, 2017-18 QUIZ 4 Solution

### **Quiz Guidelines:**

Please read the following quiz guidelines carefully:

- For this quiz, you are to work ALONE. You may use your course notes and lab materials to help answer the questions.
- Submit your answers to Courseplus by 5pm Friday March 9<sup>th</sup>
- DO NOT discuss this quiz or your solution to this quiz with other students from the course starting Wednesday March 7<sup>th</sup> through Friday March 9<sup>th</sup>. The solution to the quiz will be available Saturday, March 10<sup>th</sup>.
- By submitting your answers to Courseplus, you are acknowledging that you have read the guidelines carefully and will adhere to these guidelines.

### **Scientific Background:**

This quiz is an analysis of data from the Supplemental Nutrition Assistance Program (SNAP) which is the new name for the federal Food Stamp Program. Data include county-level population size and SNAP enrollees for 3,205 US counties for years 2000 to 2010. The population size and SNAP enrollees are based on available data from July of the corresponding year. The outcome of interest is the proportion of the county population that is enrolled in the SNAP. In addition, a metro vs. non-metro indicator for each county is provided. There is no missing data.

#### **Notation:**

 $Y_{ij}$  is the proportion enrolled in SNAP for county i (i in 1, 2, ..., 3205) during year j (j = 200, 2001, ..., 2010), this value ranges between 0 and 100.

 $year_{ij}$  is the year of observation; 2000 to 2010

 $metro_i$  is the indicator that county i is a metro county (1) vs. a non-metro county (0)

## **Objectives:**

- **PRIMARY OBJECTIVE:** to estimate the yearly trend in SNAP enrollment separately for metro and non-metro counties and determine if the recent recession (starting 2007) had a different effect on SNAP enrollment for metro vs. non-metro counties.
- **SECONDARY OBJECTIVE:** summarize the variation in the growth rates of SNAP enrollment across the counties.

Figures 1 and 2 below summarize the data across years and metro/non-metro designation.

Figure 1: Distribution of the proportion of county population enrolled in SNAP by year (2000 to 2010) and metro/non-metro designation.

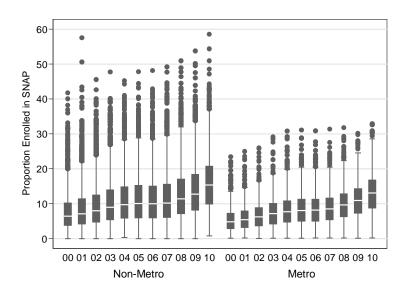
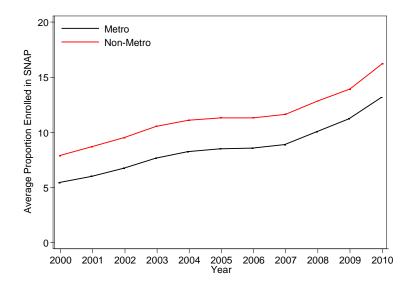


Figure 2: Average proportion enrolled in SNAP among metro/non-metro counties for years 2000 to 2010



**Linear Mixed Model specification:** To address the primary and secondary objectives of the study, we will fit a linear mixed model that allows for a county-specific estimate for the proportion enrolled in SNAP based on the county type (metro/non-metro county) and time (linear spline with knot at 2007). The county-specific estimates will include a random intercept and random slopes for both the linear and spline terms in the model. The model is given below:

$$Y_{ij} = \beta_{0i} + \beta_{1i} (year_{ij} - 2007) + \beta_{2i} (year_{ij} - 2007)^{+} + \beta_{3} metro_{i} + \beta_{4} metro_{i} \times (year_{ij} - 2007) + \beta_{5} metro_{i} \times (year_{ij} - 2007)^{+} + \varepsilon_{ij}$$

where  $(year_{ij} - 2007)^+ = 0$  if  $year_{ij} - 2007 < 0$  and  $(year_{ij} - 2007)$  if  $year_{ij} - 2007 \ge 0$ ,  $metro_i = 1$  if county i is a metro county and 0 if county i is a non-metro county.

The random effects are defined as follows:

$$\beta_{0i} = \beta_0 + b_{0i} 
\beta_{1i} = \beta_1 + b_{1i} 
\beta_{2i} = \beta_2 + b_{2i}$$

$$\begin{pmatrix} b_{0i} \\ b_{1i} \\ b_{2i} \end{pmatrix} \sim MVN \left( \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \begin{bmatrix} \sigma_0^2 & \sigma_{01} & \sigma_{02} \\ \sigma_{01} & \sigma_1^2 & \sigma_{12} \\ \sigma_{02} & \sigma_{21} & \sigma_2^2 \end{bmatrix} \right)$$

And we make the following independence assumptions for the within subject residuals and random effects.

$$\varepsilon_{ij} \sim Independent\ N(0, \sigma^2), Cov(b_{0i}, \varepsilon_{ij}) = 0, Cov(b_{1i}, \varepsilon_{ij}) = 0, Cov(b_{2i}, \varepsilon_{ij}) = 0$$

# This model was fit using mixed in STATA and the following output was obtained:

Mixed-effects ML regression Group variable: id				Number of Number of		
			C	)bs per g	group: min =	4
					avg =	10.8
					max =	11
			V	Wald chi2	2(5) =	9607.74
Log likelihood =		F	Prob > ch	ni2 =	0.0000	
y_100	   Coef.	Std. Err.	 z	P> z	 95% Co	nf. Interval]
	, +					
time_c time_sp	!	.0081568 .0406965	69.61 15.35	0.000	.551839 .544844	
metro	:	.2764718	-10.17	0.000	-3.35426	
mecro	-2.012391	.2/04/10	-10.17	0.000	-3.33420	0 -2.270310
metro#c.time_c		0455544			254226	
1	0201355 	.0157611	-1.28	0.201	051026	7 .0107556
metro#c.time_sp	j					
1	.1938918	.0785994	2.47	0.014	.039839	7 .3479438
_cons	12.24922	.1430656	85.62	0.000	11.9688	2 12.52963
Random-effects Parameters		Estimate	Std.	Err.	[95% Conf.	Interval]
id: Unstructured						
sd(time_c)		.3583502	.0053946		.3479315	.3690809
sd(time_sp)		1.601304	.0303115		1.542984	1.66183
	sd(_cons)	6.825124	.0869	9663	6.656784	6.997721
corr(time_c,time_sp)		2788328			3201314	2364764
corr(time_c,_cons)		.6696656			.6472047	.6909649
corr(time_sp,_cons)		2080311 -+	.0208	3766 	2485633 	1667717
:	sd(Residual)	1.031977	.0046	5761	1.022853	1.041183
LR test vs. linear regression: chi2(6) = 94931.65 Prob > chi2 = 0.0000						
. lincom time_c + 1.metro#c.time_c						
$(1) [y_100]time_c + [y_100]1.metro#c.time_c = 0$						
y_100	Coef. S	Std. Err.	z F	 > z	[95% Conf.	Interval]
(1)	.5480702	.0134278 40	 0.82 0	0.000	.5217521	.5743883

- 1. The model we described and fit above assumes that
  - a.  $Var(Y_{ij})$  is constant with time
  - b.  $Var(Y_{ij})$  is constant with county type (metro vs. non-metro)
  - c.  $Var(Y_{ij})$  differs with time
  - d.  $Var(Y_{ij})$  differs by county type (metro vs. non-metro)
- 2. The estimated mean proportion enrolled in SNAP in 2007 in non-metro counties is 12.25 (SE: 0.14). The estimated standard deviation of the random intercept is 6.83 (SE: 0.09). Which of the following is the most accurate interpretation of these two values?
  - a. Roughly 95% of non-metro counties in 2007 have expected mean proportion enrolled in SNAP ranging from 12.25 +/- 2 (0.14)
  - b. Roughly 95% of non-metro counties in 2007 have expected mean proportion enrolled in SNAP ranging from 12.25 +/- 2 (6.83)
  - c. Roughly 95% of non-metro counties in 2007 have expected mean proportion enrolled in SNAP ranging from 12.25 +/- 2 (0.09)
  - d. Roughly 95% of non-metro counties in 2007 have expected mean proportion enrolled in SNAP ranging from 6.83 + /- 2 (0.09)
- 3. Prior to 2007, the proportion enrolled in SNAP among metro counties increases by 0.55% per year (95% CI: 0.52 to 0.57). There is considerable variation in the growth rates in SNAP enrollment across counties. The estimate of the standard deviation in the linear growth rate prior to 2007 in metro counties is:
  - a. 0.358 (SE: 0.005)
  - b. 1.601 (SE: 0.030)
  - c. 6.825 (SE: 0.087)
  - d. 0.670 (SE: 0.011)
- 4. We estimate that the correlation between the random intercept  $(b_{0i})$  and the linear random slope  $(b_{1i})$  is 0.67. This correlation can be interpreted as:
  - a. Counties with higher than average growth in enrollment in SNAP prior to 2007 will tend to have higher than average enrollment in SNAP in 2007.
  - b. Counties with higher than average growth in enrollment in SNAP prior to 2007 will tend to have lower than average enrollment in SNAP in 2007.
  - c. Counties with lower than average growth in enrollment in SNAP prior to 2007 will tend to have higher than average enrollment in SNAP in 2007.
  - d. This value is not important as it is not statistically significantly different from 0 and therefore, these two random effects are independent.

5. Suppose now that counties receive additional federal revenue to support social worker and child healthcare outreach if the proportion of the county population enrolled in SNAP exceeds 15 percent. NOTE: I'm making this up!

You specified the random intercept logistic model below:

$$log\left[\frac{P(Y_{ij} > 15)}{P(Y_{ij} \le 15)}\right] = \beta_{0i} + \beta_1 (year_{ij} - 2007) + \beta_2 (year_{ij} - 2007)^+ + \beta_3 metro_i + \beta_4 metro_i \times (year_{ij} - 2007) + \beta_5 metro_i \times (year_{ij} - 2007)^+$$

where  $(year_{ij} - 2007)^+ = 0$  if  $year_{ij} - 2007 < 0$  and  $(year_{ij} - 2007)$  if  $year_{ij} - 2007 \ge 0$ ,  $metro_i = 1$  if county i is a metro county and 0 if county i is a non-metro county.

The random effects are defined as follows:  $\beta_{0i} = \beta_0 + b_{0i}, b_{0i} \sim N(0, \sigma_0^2)$ 

The interpretation for  $\beta_1$  is:

- a. The yearly change in the log odds of greater than 15 percent of the population enrolled in SNAP during 2000 to 2007 among all counties included in the study.
- b. The yearly change in the log odds of greater than 15 percent of the population enrolled in SNAP during 2000 to 2007 for a given/specific nonmetro county.
- c. The odds of greater than 15 percent of the population enrolled in SNAP in 2007 among all counties included in the study.
- d. The odds of greater than 15 percent of the population enrolled in SNAP in 2007 for the "average" non-metro county.
- 6. If you were to add a random slope for  $(year_{ij} 2007)$  to the random intercept logistic regression model above; i.e. define  $\beta_{1i} = \beta_1 + b_{1i}$ ,  $b_{1i} \sim N(0, \sigma_1^2)$ . The interpretation for  $\beta_1$ :
  - a. cannot be determined by the information provided.
  - b. remains the same; adding a random slope to the random intercept logistic regression model does not change any of the interpretations.
  - c. changes;  $\beta_1$  now represents the yearly change in the log odds of greater than 15 percent of the population enrolled in SNAP during 2000 to 2007 for the nonmetro county with "average" yearly change.