

**Stat 421: Applied Categorical Data Analysis, Spring 2020**

**Lec. 002**

**Derek Bean**

**Homework 9**

**Due: Friday 4/24/20 by 11:59pm CST through Canvas**

Each question worth 10 points. TOTAL: 50 points.

**Suggested Readings in Agresti (2nd Ed.)**

1. Chapter 4, Section 4.3: §4.3.1–4.3.3
2. Chapter 4, Section 4.4

**Problems**

1. A study in Florida found that the death penalty was given in 19 out of 151 cases in which a white killed a white, in 0 out of 9 cases in which white killed a black, in 11 out of 63 cases in which a black killed a white, and in 6 out of 103 cases in which a black killed a black. The below table shows software output when logic model is fit for death penalty as a response (1 = yes), with defendant's race (1 = white) and victim's race (1 = white) as indicator predictors.

Parameter	Estimate	Standard Error	Likelihood Ratio		Chi-Square
			95% Conf. Limits		
Intercept	−3.5961	0.5069	−4.7754	−2.7349	50.33
def	−0.8678	0.3671	−1.5633	−0.1140	5.59
vic	2.4044	0.6006	1.3068	3.7175	16.03
LR Statistics					
	Source	DF	Chi-Square	Pr > ChiSq	
	def	1	5.01	0.0251	
	vic	1	20.35	<.0001	

- (a) Based on the parameter estimates, which group is most likely to have the “yes” response? Estimate the probability in that case.

- (b) Interpret the parameter estimate for victim's race.
- (c) Using the information shown, construct and interpret a 95% likelihood-ratio confidence interval for the conditional odds ratio between death penalty verdict and victim's race.
- (d) Test the effect of victim's race, controlling for defendant's race, using a Wald test or likelihood-ratio test. Interpret.

2. The below table shows results of an eight-center clinical trial to compare a drug to placebo for curing an infection. At each center, subjects were randomly assigned to groups.

Center	Treatment	Response		Sample Odds Ratio
		Success	Failure	
1	Drug	11	25	1.19
	Control	10	27	
2	Drug	16	4	1.82
	Control	22	10	
3	Drug	14	5	4.80
	Control	7	12	
4	Drug	2	14	2.29
	Control	1	16	
5	Drug	6	11	$\infty$
	Control	0	12	
6	Drug	1	10	$\infty$
	Control	0	10	
7	Drug	1	4	2.0
	Control	1	8	
8	Drug	4	2	0.33
	Control	6	1	

Source: P. J. Beitler and J. R. Landis, *Biometrics*, **41**: 991–1000, 1985.

- (a) Analyze these data using logistic regression, describing and making inferences about the group effect.

- (b) Give possible reasons for why we would want to control for the center variable in our model.
3. A sample of subjects were asked their opinion about current laws legalizing abortion (support, oppose). For the explanatory variables gender  $G$  (female, male), religious affiliation  $R$  (Protestant, Catholic, Jewish), and political party affiliation  $P$  (Democrat, Republican, Independent), the model for the probability  $\pi$  of supporting legalized abortion,

$$\text{logit}(\pi) = \alpha + \beta_h^G + \beta_i^R + \beta_j^P,$$

has reported parameter estimates (setting the parameter for the last listed category of a variable equal to 0)  $\hat{\alpha} = -0.11$ ,  $\hat{\beta}_1^G = 0.16$ ,  $\hat{\beta}_2^G = 0$ ,  $\hat{\beta}_1^R = -0.57$ ,  $\hat{\beta}_2^R = -0.66$ ,  $\hat{\beta}_3^R = 0$ ,  $\hat{\beta}_1^P = 0.84$ ,  $\hat{\beta}_2^P = -1.67$ ,  $\hat{\beta}_3^P = 0$ .

- (a) Interpret how the odds of supporting legalized abortion depend on gender.
- (b) Find the estimated probability of supporting legalized abortion for (i) male Catholic Republicans and (ii) female Jewish Democrats.
- (c) If we defined parameters such that the *first* category of a variable has value 0, then what would  $\hat{\beta}^G$  equal? Show then how to obtain the odds ratio that describes the conditional effect of gender.
- (d) If we defined parameters such that they sum to 0 across the categories of a variable, then what would  $\hat{\beta}_1^G$  and  $\hat{\beta}_2^G$  equal? Show then how to obtain the odds ratio that describes the conditional effect of gender.
4. The table below shows estimated effects for a fitted logistic regression model with squamous cell esophageal cancer (1 = yes, 0 = no) as the response variable  $Y$ . Smoking status ( $S$ ) equals 1 for at least one pack per day and 0 otherwise, alcohol consumption ( $A$ ) equals the average number of alcoholic drinks consumed per day, and race ( $R$ ) equals 1 for blacks and 0 for whites.

Variable	Effect	<i>P</i> -value
Intercept	−7.00	<0.01
Alcohol use	0.10	0.03
Smoking	1.20	<0.01
Race	0.30	0.02
Race × smoking	0.20	0.04

- (a) To describe the race-by-smoking interaction, construct the prediction equation when  $R = 1$  and again when  $R = 0$ . Find the fitted  $YS$  conditional odds ratio for each case. Similarly, construct the prediction equation when  $S = 1$  and again when  $S = 0$ , and find the fitted  $YR$  conditional odds ratio for each case. Note that, for each association, the coefficient of the cross-product term is the difference between the log odds ratios at the two fixed levels for the other variable.
- (b) Explain what the coefficients of  $R$  and  $S$  represent for the coding as given above. What hypotheses do the  $P$ -values refer to for these variables?
- (c) Suppose the model also contained an  $A \times R$  interaction term, with coefficient 0.04. In the prediction equation, show that this represents the difference between the effect of  $A$  for blacks and for whites.
5. The table below shows results of a study about  $Y$  = whether a patient having surgery with general anesthesia experienced a sore throat on waking (1 = yes) as a function of  $D$  = duration of the surgery (minutes) and  $T$  = type of device used to secure the airway (0 = laryngeal mask airway, 1 = tracheal tube).

Patient	$D$	$T$	$Y$	Patient	$D$	$T$	$Y$	Patient	$D$	$T$	$Y$
1	45	0	0	13	50	1	0	25	20	1	0
2	15	0	0	14	75	1	1	26	45	0	1
3	40	0	1	15	30	0	0	27	15	1	0
4	83	1	1	16	25	0	1	28	25	0	1
5	90	1	1	17	20	1	0	29	15	1	0
6	25	1	1	18	60	1	1	30	30	0	1
7	35	0	1	19	70	1	1	31	40	0	1
8	65	0	1	20	30	0	1	32	15	1	0
9	95	0	1	21	60	0	1	33	135	1	1
10	35	0	1	22	61	0	0	34	20	1	0
11	75	0	1	23	65	0	1	35	40	1	0
12	45	1	1	24	15	1	0				

Source: Data from D. Collett, in *Encyclopedia of Biostatistics*, Wiley, New York, 1998, pp. 350–358. Predictors are  $D$  = duration of surgery,  $T$  = type of device.

- Fit a main effects model using these predictors. Interpret parameter estimates.
- Conduct inference about the  $D$  effect in the model in part (a).
- Fit a model permitting interaction. Report the prediction equation for the effect of  $D$  when (i)  $T = 1$ , (ii)  $T = 0$ . Interpret.
- Conduct inference about whether you need the interaction term in the model in (c).