Abstract

When dependent variables are ordinal rather than continuous, conventional OLS regression techniques are inappropriate. The ordered logit model, also known as the proportional odds model, is a popular method in such cases. However, in many instances, generalized ordered logit (gologit) models may be a superior alternative. Gologit models can be less restrictive than proportional odds models, whose assumptions are often violated, and more parsimonious than methods like multinomial logit that ignore the ordering of categories altogether. At the same time, the gologit model offers challenges of its own with regards to proper usage and interpretation. In this paper, we discuss the rationale behind the gologit model and show how it can be estimated using the gologit2 routine in Stata. We also discuss potential problems that are possible.

The Ordered Logit / Proportional Odds Model

Long and Freese (2006) present data from the 1977/1989 General Social Survey. Respondents are asked to evaluate the following statement: "A working mother can establish just as warm and secure a relationship with her child as a mother who does not work." Responses were coded as 1 = Strongly Disagree (1SD), 2 = Disagree (2D), 3 = Agree (3A), and 4 = Strongly Agree (4SA). Explanatory variables are yr89 (survey year; 0 = 1977, 1 = 1989), male (0 = female, 1 = male), white (0 = nonwhite, 1 = white), age (measured in years), ed (years of education), and prst (occupational prestige scale). Stata's ologit yields the following results.

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. use "http://www.indiana.edu/~jslsoc/stata/spex data/ordwarm2.dta"
(77 & 89 General Social Survey)
. * Ordered logit model
. ologit warm yr89 male white age ed prst, nolog
                                                      Number of obs = 2293
LR chi2(6) = 301.72
Ordered logistic regression
                                                      Prob > chi2
Pseudo B2
                                                                       = 0.0000
Log likelihood = -2844.9123
                                                                       =
                                                       Pseudo R2
                                                                              0.0504
_____
        warm | Coef. Std. Err. z P>|z| [95% Conf. Interval]
yr89 |.5239025.07989886.560.000.3673037.6805013male |-.7332997.0784827-9.340.000-.8871229-.5794766white |-.3911595.1183808-3.300.001-.6231815-.1591374age |-.0216655.0024683-8.780.000-.0265032-.0168278ed |.0671728.0159754.200.000.0358624.0984831prst |.0060727.00329291.840.065-.0003813.0125267
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Generalized Ordered Logit Models, Richard Williams, 2010 MSS Meetings - p. 1

+-				
/cut1	-2.465362	.2389126	-2.933622	-1.997102
/cut2	630904	.2333155	-1.088194	173614
/cut3	1.261854	.2340179	.8031873	1.720521

These results are relatively straightforward, intuitive and easy to interpret. People tended to be more supportive of working mothers in 1989 than in 1977. Males, whites and older people tended to be less supportive of working mothers, while better educated people and people with higher occupational prestige were more supportive.

But, while the results may be straightforward, intuitive, and easy to interpret, are they correct? The brant command (part of Long and Freese's spost routines) indicates that the assumptions of the ordered logit/ proportional odds model are not met.

	y>1	y>2	y>3
yr89	.9647422	.56540626	.31907316
male	30536425	69054232	-1.0837888
white	55265759	31427081	39299842
age	0164704	02533448	01859051
ed	.10479624	.05285265	.05755466
prst	00141118	.00953216	.00553043
_cons	1.8584045	.73032873	-1.0245168

Brant Test of Parallel Regression Assumption

Variable		chi2	p>chi2	df
All		49.18	0.000	12
yr89 male white age ed prst	 	13.01 22.24 1.27 7.38 4.31 4.33	0.001 0.000 0.531 0.025 0.116 0.115	2 2 2 2 2 2 2 2

A significant test statistic provides evidence that the parallel regression assumption has been violated.

Proportional Odds & Partial Proportional Odds/ Parallel Lines & Non-Parallel Lines

Model 0: Perfect Propo	ortional Odds/ Pa	rallel Line	S			
	attitu	ude				
gender .	SD D	A	SA	Tota	1	
Male 25	50 250	250	250	1,00	00	
Female 10)0 150	250	500	1,00	00	
Total 3	50 400	500	750	2,00	00	
	1 versus 2, 3, 4		1 & 2 versus 3	& 4	1, 2, 3 versus 4	
Odds _M	750/250 = 3		500/500 = 1		250/750 = 1/3	
Odds _F	900/100 = 9		750/250 = 3		500/500 = 1	
OR (Odds _F / Odds _M)	9/3 = 3		3/1 = 3		1/(1/3) = 3	
Gologit2 Betas	1.098612		1.098612		1.098612	
Gologit2 χ^2 (3 d.f.)	176.63 (p = 0.000	0)				
Ologit χ^2 (1 d.f.)	176.63 (p = 0.000	(00				
Ologit Beta (OR) 1.098612 (3.00)						
Brant Test (2 d.f.)	0.0 (p = 1.000)					
Comment	If proportional odds holds, then the odds ratios should be the same for each of the ordered					
	dichotomizations of the dependent variable. Proportional Odds works perfectly in this model, as the odds ratios are all 3. Also, the Betas are all the same, as they should be.					

Model 1: Partial Proportional Odds I								
		attitud	2					
gender		D	A			1		
 Male		250		1		0		
Female +	100	300	300	300	1,00	0		
 Total								
	1 ve	rsus 2, 3, 4	1 & 2	2 versus 3 & 4	4	1, 2, 3 versus 4		
Odds _M	750/250 = 3		500/	500 = 1		250/750 = 1/3		
Odds _F	900/1	00 = 9	600/-	400 = 1.5		300/700 = 3/7		
OR (Odds _F / Odds _M)	9/3 =	3	1.5/1	= 1.5		(3/7)/(1/3) = 1.28		
Gologit2 Betas	1.098	3612	.4054	4651		.2513144		
	•							
Gologit2 χ^2 (3 d.f.)	80.07	(p = 0.0000)						
Ologit χ^2 (1 d.f.)								
Ologit Beta (OR) .4869136 (1.627286))					
Brant Test (2 d.f.)	· ·							
Comment	Gender has its greatest effect at the lowest levels of attitudes, i.e. women are much less likely to					women are much less likely to		
	strongly disagree than men are, but other differences are smaller. The effect of gender is							
	consi	stently positive, i	.e. the difference	es involve magni	tude, not s	gn.		

Model 2: Partial	Proport	ional Odds II				
	порон					
		attitude	5			
gender	S	D D	A	SA	Tota	1
		250				
Female	10	400	250	250	1,00	0
+ Total) 650		500		0
		1 versus 2, 3, 4	1	& 2 versus	3 & 4	1, 2 3 versus 4
Odds _M			5	00/500 = 1		250/750 = 1/3
Odds _F		900/100 = 9		00/500 = 1		250/750 = 1/3
OR (Odds _F / Odds	OR (Odds _F / Odds _M)		1/	1/1 = 1 $(1/3)/(1/3) = 1$		(1/3)/(1/3) = 1
Gologit2 Betas		1.098612	0	0		0
Gologit2 χ^2 (3 d.f.))	101.34 (p = 0.0000)				
Ologit χ^2 (1 d.f.)		9.13 (p = 0.0025)				
Ologit Beta (OR)		.243576 (1.275803)				
Brant Test (2 d.f.)		83.05 (p = 0.000)				
Comment		Gender has its greatest – and only – effect at the lowest levels of attitudes, i.e. women are much less likely to strongly disagree than men are. But, this occurs entirely because they are much more likely to disagree rather than strongly disagree. Other than that, there is no gender effect; men and women are equally likely to agree and to strongly agree. The ologit estimate underestimates the effect of gender on the lower levels of attitudes and overestimates its effect at the higher levels.				

Model 3: Partial	lodel 3: Partial Proportional Odds III							
I		attitud	le					
gender	SD	D	A	SA	Total			
Male	250	250	250	+ 250	1,000			
Female	100	400	400	100	1,000			
Total	350	650	650	+ 350	2,000			

	1 versus 2, 3, 4	1 & 2 versus 3 & 4	1, 2, 3 versus 4
Odds _M	750/250 = 3	500/500 = 1	250/750 = 1/3
Odds _F	900/100 = 9	500/500 = 1	100/900 = 1/9
OR (Odds _F / Odds _M)	9/3 = 3	1/1 = 1	(1/9)/(1/3) = 1/3
Gologit2 Betas	1.098612	0	-1.098612

Gologit2 χ^2 (3 d.f.)	202.69 (p = 0.0000)
Ologit χ^2 (1 d.f.)	0.00 (p = 1.0000)
Ologit Beta (OR)	0 (1.00))
Brant Test (2 d.f.)	179.71 (p = 0.000)
Comment	The effect of gender varies in both sign and magnitude across the range of attitudes. Basically, women tend to take less extreme attitudes in either direction. They are less likely to strongly disagree than are men, but they are also less likely to strongly agree. The ologit results imply gender has no effect while the gologit results say the effect of gender is highly significant. Perhaps the current coding of attitudes is not ordinal with respect to gender, e.g. coding by intensity of attitudes rather than direction may be more appropriate. Or, suppose that, instead of attitudes, the categories represented a set of ordered hurdles, e.g. achievement levels. Women as a whole may be more likely than men to clear the lowest hurdles but less likely to clear the highest ones. If men are more variable than women, they will have more outlying cases in both directions. Use of ologit in this case would be highly misleading.

A Parsimonious Alternative: Generalized Ordered Logit/ Partial Proportional Odds

1. Unconstrained Gologit Model. All betas are free to differ across levels of j.

$$P(Y_i > j) = \frac{\exp(\alpha_j + X_i\beta_j)}{1 + [\exp(\alpha_j + X_i\beta_j)]}, j = 1, 2, ..., M - 1$$

2. Special Case: Proportional Odds. All betas the same across levels of j.

$$P(Y_i > j) = \frac{\exp(\alpha_j + X_i\beta)}{1 + [\exp(\alpha_j + X_i\beta)]}, j = 1, 2, ..., M - 1$$

3. Special Case: Partial Proportional Odds. Some betas differ across levels of j but others do not.

$$P(Y_i > j) = \frac{\exp(\alpha_j + X1_i\beta 1 + X2_i\beta 2 + X3_i\beta 3_j)}{1 + [\exp(\alpha_j + X1_i\beta 1 + X2_i\beta 2 + X3_i\beta 3_j)]}, j = 1, 2, ..., M - 1$$

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. * Partial proportional odds - relax the pl assumption when it is violated
. use "http://www.indiana.edu/~jslsoc/stata/spex data/ordwarm2.dta", clear
. gologit2 warm yr89 male white age ed prst, auto lrf store(gologit2)
  _____
Testing parallel lines assumption using the .05 level of significance...
Step 1: Constraints for parallel lines imposed for white (P Value = 0.7136)
Step 2: Constraints for parallel lines imposed for ed (P Value = 0.1589)
Step 3: Constraints for parallel lines imposed for prst (P Value = 0.2046)
Step 4: Constraints for parallel lines imposed for age (P Value = 0.0743)
Step 5: Constraints for parallel lines are not imposed for
         yr89 (P Value = 0.00093)
         male (P Value = 0.00002)
Wald test of parallel lines assumption for the final model:
 ( 1) [1SD] white - [2D] white = 0
 (2) [1SD]ed - [2D]ed = 0
 ( 3) [1SD]prst - [2D]prst = 0
 (4) [1SD]age - [2D]age = 0
 (5)
      [1SD]white - [3A]white = 0
 (6) [1SD]ed - [3A]ed = 0
 ( 7) [1SD]prst - [3A]prst = 0
 (8) [1SD]age - [3A]age = 0
          chi2(8) = 12.80
        Prob > chi2 = 0.1190
An insignificant test statistic indicates that the final model
does not violate the proportional odds/ parallel lines assumption
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If you re-estimate this exact same model with gologit2, instead of autofit you can save time by using the parameter

Genera	Generalized Ordered Logit Estimates Number of obs = LR chi2(10) =							
Log li	ikelihood	d = -2826.6182	Prob Pseud	> chi2 = lo R2 =	0.0000 0.0565			
(1) (2) (3) (4) (5) (6) (7) (8)	[1SD]ec [1SD]pr [1SD]ac [2D]whi [2D]ed [2D]prs	hite - [2D]whi A - [2D]ed = (Sst - [2D]prst ge - [2D]age = te - [3A]whit - [3A]ed = 0 St - [3A]prst e - [3A]age =	c = 0 = 0 ce = 0 = 0					
	warm	Coef.	Std. Err.	Z	P> z	[95% Conf.	Interval]	
1SD 2D	<pre>yr89 male white age ed prst _cons yr89 male white age </pre>	3328209 3832583 0216325 .0670703 .0059146 2.12173 534369 6932772 3832583 0216325	.1530091 .1275129 .1184635 .0024751 .0161311 .0033158 .2467146 .0913937 .0885898 .1184635 .0024751	6.43 -2.61 -3.24 -8.74 4.16 1.78 8.60 5.85 -7.83 -3.24 -8.74	0.000 0.009 0.001 0.000 0.000 0.074 0.000 0.000 0.000 0.001 0.000	.6837876 5827417 6154424 0264835 .0354539 0005843 1.638178 3552406 8669099 6154424 0264835	1.283572 0829002 1510742 0167814 .0986866 .0124135 2.605282 7134974 5196444 1510742 0167814	
	ed prst _cons +		.0161311 .0033158 .2358361	4.16 1.78 2.55	0.000 0.074 0.011	.0354539 0005843 .1399323	.0986866 .0124135 1.064393	
3A	yr89 male white age ed prst _cons	-1.097615 3832583 0216325 .0670703	.1125481 .1214597 .1184635 .0024751 .0161311 .0033158 .2393568	2.89 -9.04 -3.24 -8.74 4.16 1.78 -4.38	0.004 0.001 0.000 0.000 0.000 0.074 0.000	.1052197 -1.335671 6154424 0264835 .0354539 0005843 -1.517268	.5464 8595579 1510742 0167814 .0986866 .0124135 5790061	