# predict and margins

Andy Grogan-Kaylor

1 Mar 2021 08:10:44

## Background

Odds ratios, or coefficients showing the association of the independent variables with the log odds, represent the most immediate output of a logistic regression. However, for a variety of reasons, it may make sense to not only report odds ratios, but also to investigate predicted probabilities.

#### Get The Data

We start by obtaining *simulated data* from StataCorp.

- . clear all
- . graph close \_all
- . use http://www.stata-press.com/data/r15/margex, clear
  (Artificial data for margins)

#### Describe The Data

The variables are as follows:

. describe

Contains data from http://www.stata-press.com/data/r15/margex.dta
obs: 3,000 Artificial data for margins
vars: 11 27 Nov 2016 14:27

variable name	storage type	display format	value label	variable label
у	float	%6.1f		
outcome	byte	%2.0f		
sex	byte	%6.0f	sexlbl	
group	byte	%2.0f		
age	float	%3.0f		
distance	float	%6.2f		
ycn	float	%6.1f		
уc	float	%6.1f		
treatment	byte	%2.0f		
agegroup	byte	%8.0g	agelab	
arm	byte	%8.0g	ū	

Sorted by: group

## Estimate Logistic Regression (logit)

We then run a logistic regression model in which outcome is the dependent variable. sex, age and group are the independent variables.

```
. logit outcome i.sex c.age i.group, or
               log likelihood = -1366.0718
Iteration 0:
Iteration 1:
               log likelihood = -1111.4595
               log likelihood = -1069.588
Iteration 2:
Iteration 3:
               log likelihood =
Iteration 4:
               log likelihood = -1067.9941
               log likelihood = -1067.9941
Iteration 5:
Logistic regression
                                                                           3,000
                                                 Number of obs
                                                                          596.16
                                                 LR chi2(4)
                                                 Prob > chi2
                                                                          0.0000
                                                 Pseudo R2
Log likelihood = -1067.9941
                                                                          0.2182
                                                 P>|z|
     outcome
               Odds Ratio
                            Std. Err.
                                                            [95% Conf. Interval]
                                            z
         sex
                  1.64734
                              .221973
                                          3.70
                                                 0.000
                                                            1.26499
                                                                        2.145258
     female
                  1.09444
                             .0070921
                                         13.93
                                                 0.000
                                                            1.080628
                                                                        1.108429
         age
       group
                 .5568139
                             .0751806
                                         -4.34
                                                 0.000
                                                            .4273478
                                                                         .725502
          3
                 .2566074
                             .0747822
                                         -4.67
                                                 0.000
                                                            .1449462
                                                                        .4542885
                 .0038757
                             .0013558
                                        -15.87
                                                 0.000
                                                            .0019524
                                                                        .0076933
       cons
```

Note: \_cons estimates baseline odds.

### Margins (margins)

5#female

.4622129

.0303129

We use the margins command to estimate predicted probabilities at different values of sex and age.

. margins sex	, at(age =	(20 30 40 50	60))				
Predictive ma	argins : OIM		Number	of obs	=	3,000	
Expression	: Pr(outco	ome), predict(	)				
1at	: age	=	20				
2at	: age	=	30				
3at	: age	=	40				
4at	: age	=	50				
5at	: age	=	60				
		Delta-meth	od				
	Marg	in Std. Err	. z	P> z	[95%	Conf.	Interval]
at#sex							
_at#sex 1#male		34 .0031264	4.92	0.000	. 009	2657	.0215211
1#male	.01539			0.000	.009:		.0215211
1#male 1#female	.01539	.0046143	5.43	0.000	.016	0171	.0341048
1#male 1#female 2#male	.01539 .02506 .03696	.0046143 .0054588	5.43 6.77	0.000	.016	0171 2635	.0341048
1#male 1#female 2#male 2#female	.01539 .02506 .03696 .05921	.0046143 626 .0054588 .51 .0072711	5.43 6.77 8.14	0.000 0.000 0.000	.0160 .0263 .0449	0171 2635 9639	.0341048 .0476616 .0734663
1#male 1#female 2#male	.01539 .02506 .03696	.0046143 .0054588 .51 .0072711 .77 .0088815	5.43 6.77 8.14 9.65	0.000	.016	0171 2635 9639 2603	.0341048
1#male 1#female 2#male 2#female 3#male 3#female	.01539 .02506 .03696 .05921 .08566	.009 .0046143 .226 .0054588 .51 .0072711 .77 .0088815 .88 .0097333	5.43 6.77 8.14 9.65 13.62	0.000 0.000 0.000 0.000	.0160 .0262 .0449 .0682	0171 2635 9639 2603 4919	.0341048 .0476616 .0734663 .1030751 .1516458
1#male 1#female 2#male 2#female 3#male	.01539 .02506 .03696 .05921	.009 .0046143 .026 .0054588 .51 .0072711 .77 .0088815 .88 .0097333 .78 .015461	5.43 6.77 8.14 9.65	0.000 0.000 0.000 0.000	.0160 .0263 .0449 .0683	0171 2635 9639 2603 4919 1547	.0341048 .0476616 .0734663 .1030751

15.25

0.000

.4028007

.5216251

### Plotting Margins (marginsplot)

margins provides a lot of results, which can be difficult to understand. Therefore, we use marginsplot to plot these margins results. The key command is marginsplot, which could be used on its own. I have simply added the Michigan graph scheme, as well as some options to improve the graphic design of the plot.

```
. marginsplot, scheme(michigan)
   Variables that uniquely identify margins: age sex
. graph export mymarginsplot.png, width(500) replace
(file /Users/agrogan/Desktop/newstuff/categorical/predict-and-margins/mymarginsplot.png written in PNG
> format)
```

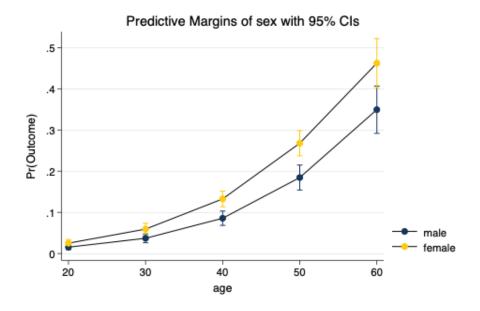


Figure 1: Graph of Predicted Margins

### Predicted Probabilities (predict)

Predicted probabilities are each participant's individual predicted probability of experiencing depression based upon the independent variables included in the model. We often denote such predicted probabilities with  $\hat{y}$ 

```
. predict yhat
(option pr assumed; Pr(outcome))
```

yhat is a variable in the data, just like any other variable, and we can tabulate and graph it.

```
. twoway (lowess yhat age if sex == 0) ///
> (lowess yhat age if sex == 1), ///
> title("Predicted Probabilities of Outcome") ///
> legend(order(1 "male" 2 "female")) ///
> scheme(michigan)

. graph export mytwoway.png, width(500) replace
(file /Users/agrogan/Desktop/newstuff/categorical/predict-and-margins/mytwoway.png written in PNG form
> at)
```

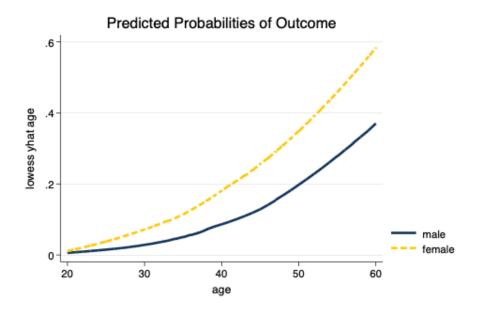


Figure 2: Bar Graph of Predicted Probabilities