

# Mixed Models for Longitudinal Ordinal and Nominal Outcomes

Don Hedeker

Department of Public Health Sciences  
Biological Sciences Division  
University of Chicago

*hedeker@uchicago.edu*

Hedeker, D. (2008). Multilevel models for ordinal and nominal variables. In J. de Leeuw & E. Meijer (Eds.), *Handbook of Multilevel Analysis*. Springer, New York.

Hedeker, D. & Gibbons, R.D. (2006). *Longitudinal Data Analysis*, chapters 10 & 11. Wiley.

Hedeker, D. (2014). Methods for multilevel ordinal data in prevention research. *Prevention Science*.

This work was supported by National Institute of Mental Health Contract N44MH32056.

## Why analyze as ordinal?

- Efficiency: Armstrong & Sloan (1989, Amer Jrn of Epid) and Strömberg (1996, Amer Jrn of Epid) report efficiency losses between 49% to 87% when dichotomizing an ordinal outcome with five categories.
- Bias: continuous model can yield correlated residuals and regressors when used for ordinal outcomes; continuous model does not take into account the ceiling and floor effects of the ordinal outcome. Results in biased estimates of regression coefficients and is most critical when the ordinal variables is highly skewed (see Bauer & Sterba, 2011, Psych Methods)
- Logic: continuous model can yield predicted values outside of the range of the ordinal variable.

# Ordinal Logistic Regression Model (aka Proportional Odds or Cumulative Logit Model) - McCullagh (1980)

$$\log \left[ \frac{P(Y \leq c)}{1 - P(Y \leq c)} \right] = \gamma_c - \mathbf{x}' \boldsymbol{\beta}$$

$c = 1, \dots, C - 1$  for the  $C$  categories of the ordinal outcome

$\mathbf{x}$  = vector of explanatory variables (plus the intercept)

$\gamma_c$  = threshold parameters; reflect cumulative logits when  $\mathbf{x} = 0$   
(for identification:  $\gamma_1 = 0$  or  $\beta_0 = 0$ )

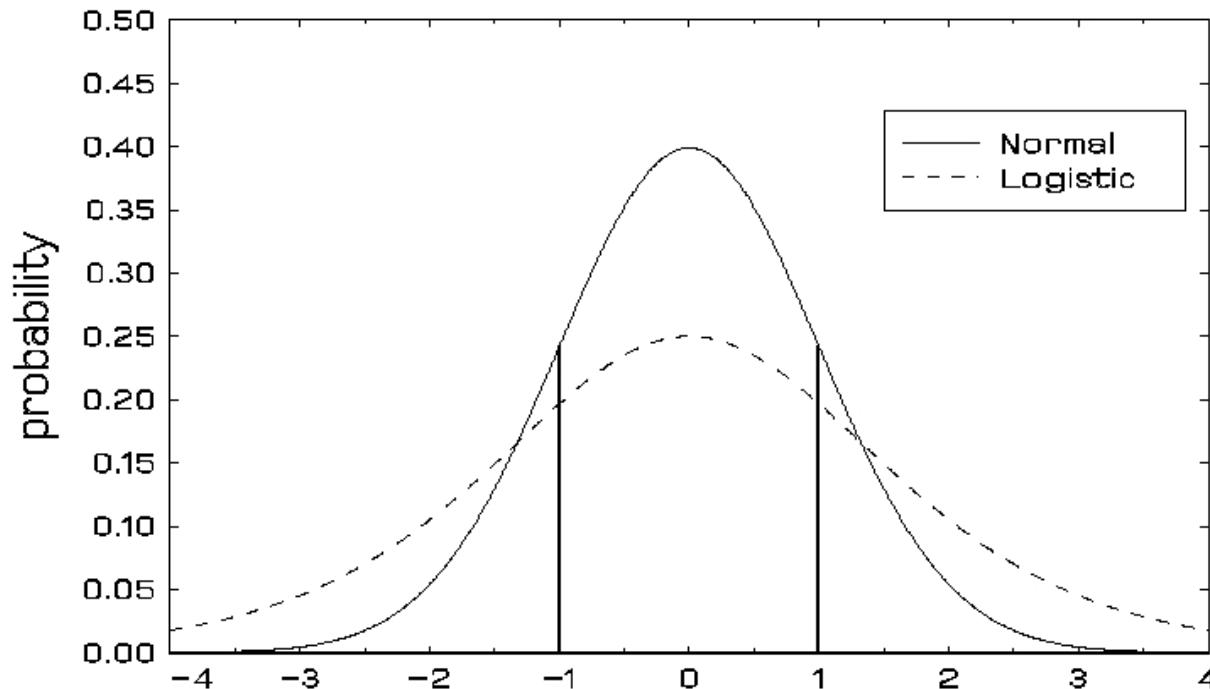
- positive association between explanatory variable  $x$  and ordinal outcome variable  $Y$  is reflected by  $\beta$
- $x$  is assumed to have the same effect on each cumulative logit (proportional odds assumption)

# Ordinal Response and Threshold Concept

Continuous  $y_i$  - unobservable latent variable - related to ordinal response  $Y_i$  via “threshold concept”

- threshold values  $\gamma_1, \gamma_2, \dots, \gamma_{C-1}$  ( $\gamma_0 = -\infty$  and  $\gamma_C = \infty$ )
- $C$  = number of ordered categories

Response occurs in category  $c$ ,  $Y_i = c$  if  $\gamma_{c-1} < y_i < \gamma_c$



# The Threshold Concept in Practice

“How was your day?”

(what is your level of satisfaction today?)

- Satisfaction may be continuous, but we sometimes emit an ordinal response:



**Great Day!**



**a day ...**



**\*?!\*\*!?** day

## Model for Latent Continuous Responses

Consider the model with  $p$  covariates for the latent response strength  $y_i$  ( $i = 1, 2, \dots, N$ ):

$$y_i = \mathbf{x}'_i \boldsymbol{\beta} + \varepsilon_i$$

- probit:  $\varepsilon_i \sim$  standard normal (mean=0, variance=1)
- logistic:  $\varepsilon_i \sim$  standard logistic (mean=0, variance= $\pi^2/3$ )

$\Rightarrow \boldsymbol{\beta}$  estimates from logistic regression are larger (in abs. value) than from probit regression by approximately  $\sqrt{\pi^2/3} = 1.8$

Underlying latent variable

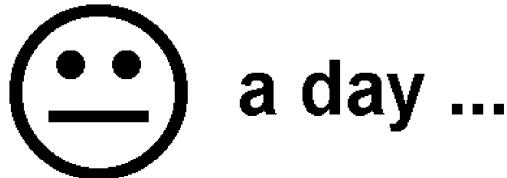
- useful way of thinking of the problem
- not an essential assumption of the model

# Mixed-effects ordinal logistic regression model

(Hedeker & Gibbons, 1994, 1996)

- $i = 1, \dots, N$  level-2 units (clusters or subjects)
- $j = 1, \dots, n_i$  level-1 units (subjects or repeated observations)
- $c = 1, 2, \dots, C$  response categories
- $Y_{ij}$  = ordinal response of level-2 unit  $i$  and level-1 unit  $j$

How was your day? (asked repeatedly each day for a week)



## Mixed-effects ordinal logistic regression model

$$\lambda_{ijc} = \log \left[ \frac{P_{ijc}}{(1 - P_{ijc})} \right] = \gamma_c - (\boldsymbol{x}'_{ij} \boldsymbol{\beta} + \boldsymbol{z}'_{ij} \boldsymbol{\nu}_i)$$

- $P_{ijc} = \Pr(Y_{ij} \leq c \mid \boldsymbol{\nu}; \gamma_c, \boldsymbol{\beta}, \boldsymbol{\Sigma}_{\boldsymbol{\nu}}) = \frac{1}{1+\exp(-\lambda_{ijc})}$
- $p_{ijc} = \Pr(Y_{ij} = c \mid \boldsymbol{\nu}; \gamma_c, \boldsymbol{\beta}, \boldsymbol{\Sigma}_{\boldsymbol{\nu}}) = P_{ijc} - P_{ijc-1}$
- $C - 1$  strictly increasing model thresholds  $\gamma_c$
- $\boldsymbol{x}_{ij} = p \times 1$  covariate vector
- $\boldsymbol{z}_{ij} = r \times 1$  design vector for random effects
- $\boldsymbol{\beta} = p \times 1$  fixed regression parameters
- $\boldsymbol{\nu}_i = r \times 1$  random effects for level-2 unit  $i \sim N(\mathbf{0}, \boldsymbol{\Sigma}_{\boldsymbol{\nu}})$

## Model for Latent Continuous Responses

Model with  $p$  covariates for the latent response strength  $y_{ij}$ :

$$y_{ij} = \mathbf{x}'_{ij}\boldsymbol{\beta} + v_{0i} + \varepsilon_{ij}$$

where  $v_{0i} \sim N(0, \sigma_v^2)$ , and assuming

- $\varepsilon_{ij} \sim$  standard normal (mean 0 and  $\sigma^2 = 1$ ) leads to mixed-effects ordinal probit regression
- $\varepsilon_{ij} \sim$  standard logistic (mean 0 and  $\sigma^2 = \pi^2/3$ ) leads to mixed-effects ordinal logistic regression

## Underlying latent variable

- not an essential assumption of the model
- useful for obtaining intra-class correlation ( $r$ )

$$r = \frac{\sigma_v^2}{\sigma_v^2 + \sigma^2}$$

and for design effect ( $d$ )

$$d = \frac{\sigma_v^2 + \sigma^2}{\sigma^2} = 1/(1 - r)$$

ratio of actual variance to the variance that would be obtained by simple random sampling (holding sample size constant)

# Scaling of regression coefficients

*Fixed-effects model*

$\beta$  estimates from logistic regression are larger (in abs. value) than from probit regression by approximately

$$\sqrt{\frac{\pi^2/3}{1}} = 1.8$$

because

- $V(y) = \sigma^2 = \pi^2/3$  for logistic
- $V(y) = \sigma^2 = 1$  for probit

## *Mixed-effects model*

$\beta$  estimates from mixed-effects (random intercepts) model are larger (in abs. value) than from fixed-effects model by approximately

$$\sqrt{d} = \sqrt{\frac{\sigma_v^2 + \sigma^2}{\sigma^2}}$$

because

- $V(y) = \sigma_v^2 + \sigma^2$  in mixed-effects (random intercepts) model
- $V(y) = \sigma^2$  in fixed-effects model
- difference depends on size of random-effects variance  $\sigma_v^2$
- more complex for models with multiple random effects

# Treatment-Related Change Across Time

Data from the NIMH Schizophrenia collaborative study on treatment related changes in overall severity. IMPS item 79, *Severity of Illness*, was scored as:

- 1 = normal or borderline mentally ill
- 2 = mildly or moderately ill
- 3 = markedly ill
- 4 = severely or among the most extremely ill

The experimental design and corresponding sample sizes:

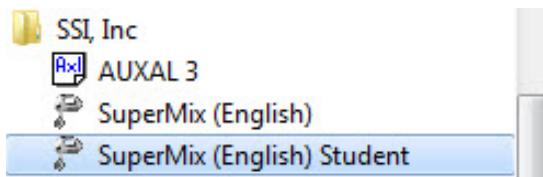
Group	Sample size at Week							<i>completers</i>
	0	1	2	3	4	5	6	
PLC (n=108)	107	105	5	87	2	2	70	65%
DRUG (n=329)	327	321	9	287	9	7	265	81%

*Drug = Chlorpromazine, Fluphenazine, or Thioridazine*

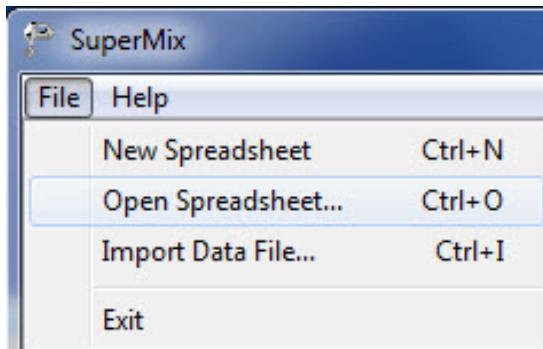
Main question of interest:

- Was there differential improvement for the drug groups relative to the control group?

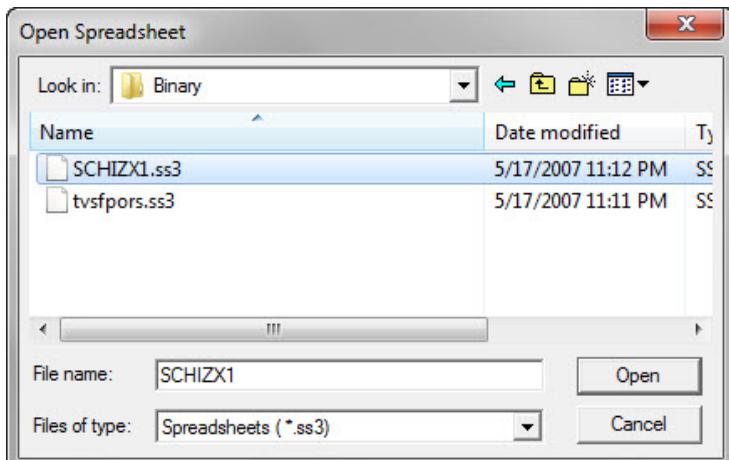
- Under SSI, Inc > “SuperMix (English)” or “SuperMix (English) Student”



- Under “File” click on “Open Spreadsheet”



- Open C:\SuperMixEn Examples\Workshop\Binary\SCHIZX1.ss3  
(or C:\SuperMixEn Student Examples\Workshop\Binary\SCHIZX1.ss3)



# C:\SuperMixEn Examples\Workshop\Binary\SCHIZX1.ss3

The screenshot shows a software window titled "SCHIZX1.ss3". The main area is a data grid with 28 rows and 8 columns. The columns are labeled (A) through (H). Row 1 contains the column headers: (A)\_Patient, (B)\_Imps79, (C)\_Imps79D, (D)\_Imps790, (E)\_TxDrug, (F)\_Week, (G)\_SqrWhee, and (H)\_Tx\*SWe. The data grid contains the following values:

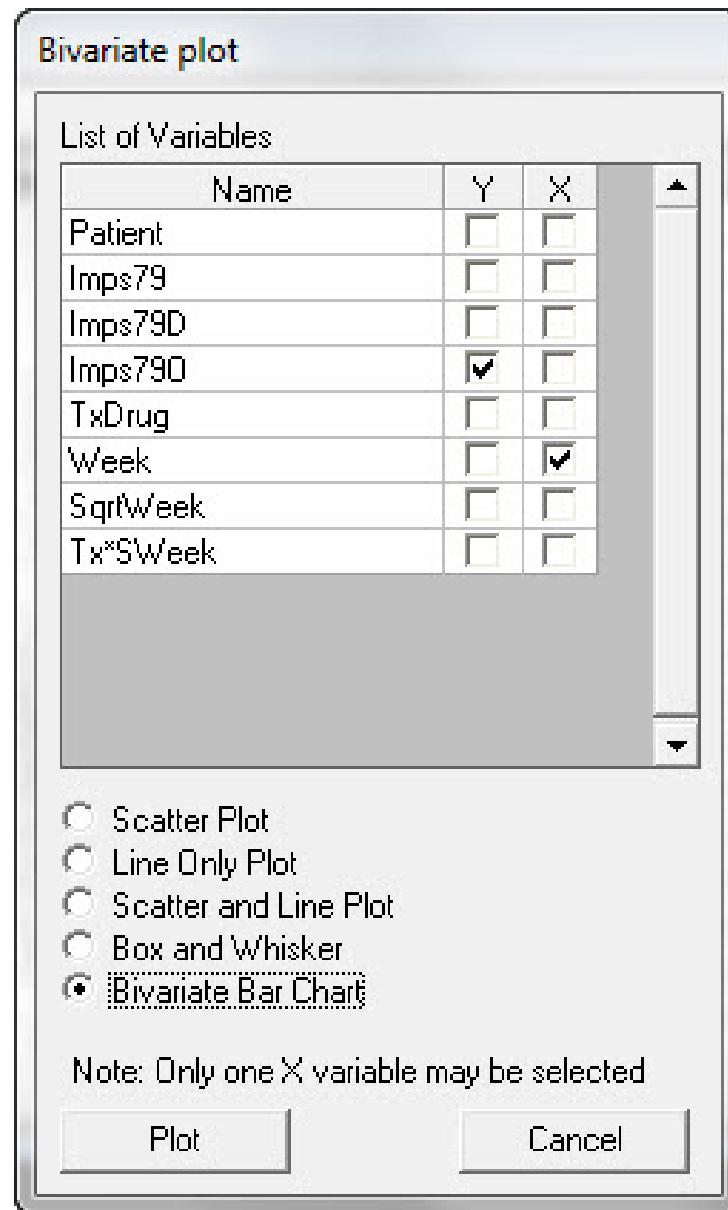
	(A)_Patient	(B)_Imps79	(C)_Imps79D	(D)_Imps790	(E)_TxDrug	(F)_Week	(G)_SqrWhee	(H)_Tx*SWe
1	1103	5.50	1	4	1	0	0.00	0.00
2	1103	3.00	0	2	1	1	1.00	1.00
3	1103	-9.00	-9	-9	1	2	1.41	1.41
4	1103	2.50	0	2	1	3	1.73	1.73
5	1103	-9.00	-9	-9	1	4	2.00	2.00
6	1103	-9.00	-9	-9	1	5	2.24	2.24
7	1103	4.00	1	2	1	6	2.45	2.45
8	1104	6.00	1	4	1	0	0.00	0.00
9	1104	3.00	0	2	1	1	1.00	1.00
10	1104	-9.00	-9	-9	1	2	1.41	1.41
11	1104	1.50	0	1	1	3	1.73	1.73
12	1104	-9.00	-9	-9	1	4	2.00	2.00
13	1104	-9.00	-9	-9	1	5	2.24	2.24
14	1104	2.50	0	2	1	6	2.45	2.45
15	1105	4.00	1	2	1	0	0.00	0.00
16	1105	3.00	0	2	1	1	1.00	1.00
17	1105	-9.00	-9	-9	1	2	1.41	1.41
18	1105	1.00	0	1	1	3	1.73	1.73
19	1105	-9.00	-9	-9	1	4	2.00	2.00
20	1105	-9.00	-9	-9	1	5	2.24	2.24
21	1105	-9.00	-9	-9	1	6	2.45	2.45
22	1106	3.00	0	2	1	0	0.00	0.00
23	1106	1.00	0	1	1	1	1.00	1.00
24	1106	-9.00	-9	-9	1	2	1.41	1.41
25	1106	1.50	0	1	1	3	1.73	1.73
26	1106	-9.00	-9	-9	1	4	2.00	2.00
27	1106	-9.00	-9	-9	1	5	2.24	2.24
28	1106	1.00	0	1	1	6	2.45	2.45

Select Imps790 column, then “Edit” > “Set Missing Value”

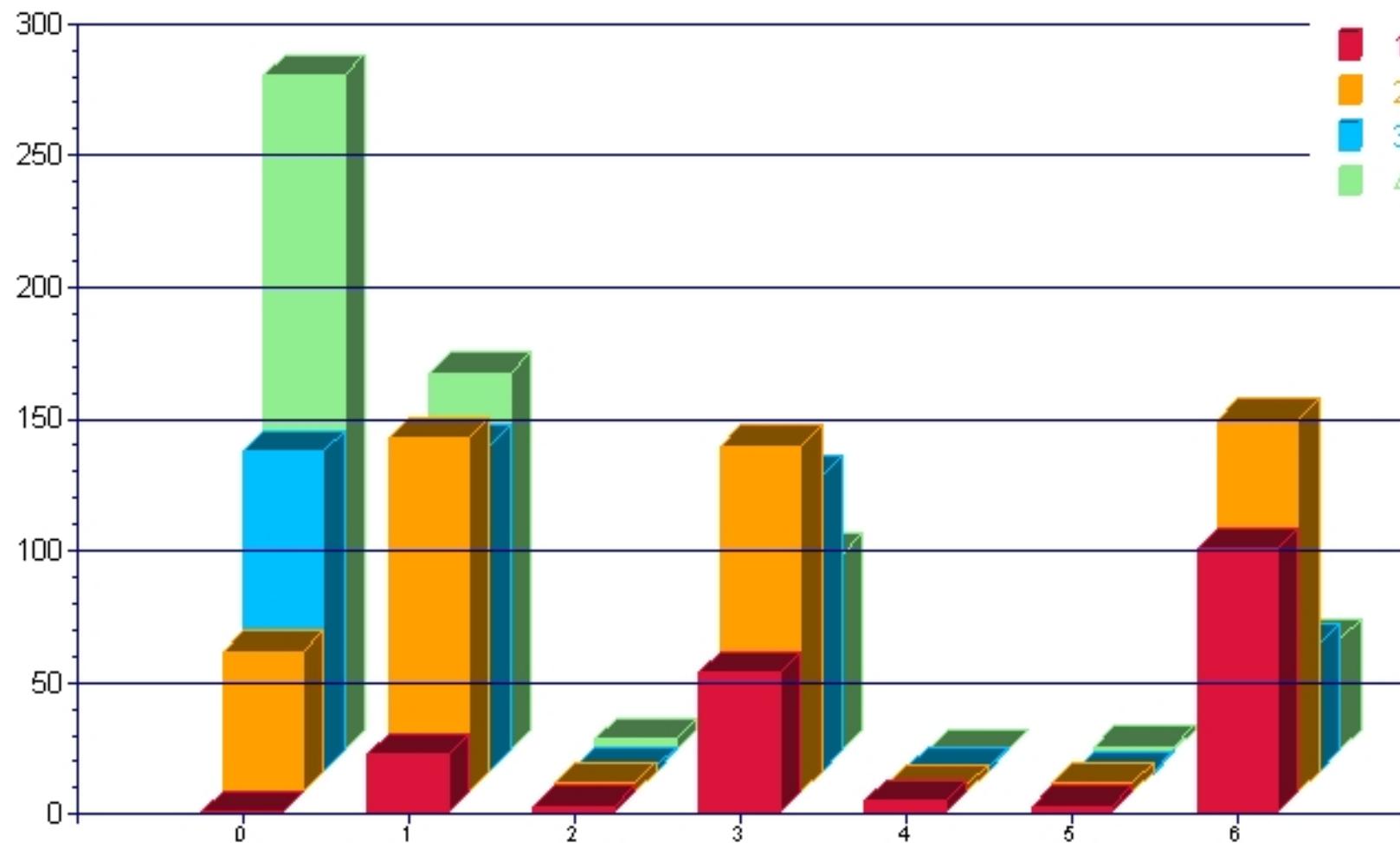
The screenshot shows the SuperMix software interface. The main window displays a data grid titled "SCHIZX1.ss3" with 19 rows and 8 columns. The columns are labeled: (A)\_Patient, (B)\_Imps79, (C)\_Imps79D, (D)\_Imps790, (E)\_TxDrug, (F)\_Week, (G)\_SqrtWhee, and (H)\_Tx\*. The row number "4" is selected. The "Imps790" column contains numerical values and a missing value code "-9". A context menu is open over the "Imps790" column, with the option "Set Missing Value" highlighted. A modal dialog box titled "Set Missing Value" is displayed, showing the "Missing Value Code:" field containing "-9". The dialog has "OK" and "Cancel" buttons.

	(A)_Patient	(B)_Imps79	(C)_Imps79D	(D)_Imps790	(E)_TxDrug	(F)_Week	(G)_SqrtWhee	(H)_Tx*
1	1103	5.50	1	4	1	0	0.00	
2	1103	3.00	0	2	1	1	1.00	
3	1103	-9.00	-9	-9	1	2	1.41	
4	1103	2.50	0	2	1	3	1.73	
5	1103	-9.00	-9	-9	1	4	2.00	
6	1103	-9.00	-9	-9	1	5	2.24	
7	1103	4.00	1	2	1	6	2.45	
8	1104	6.00	1	4	1	0	0.00	
9	1104	3.00	0	2	1	1	1.00	
10	1104	-9.00	-9	-9	1	2	1.41	
11	1104	1.50	0	1	1	3	1.73	
12	1104	-9.00	-9	-9	1	4	2.00	
13	1104	-9.00	-9	-9	1	5	2.24	
14	1104	2.50	0	2	1	6	2.45	
15	1105	4.00	1	2	1	0	0.00	
16	1105	3.00	0	2	1	1	1.00	
17	1105	-9.00	-9	-9	1	2	1.41	
18	1105	1.00	0	1	1	3	1.73	
19	1105	-9.00	-9	-9	1	4	2.00	

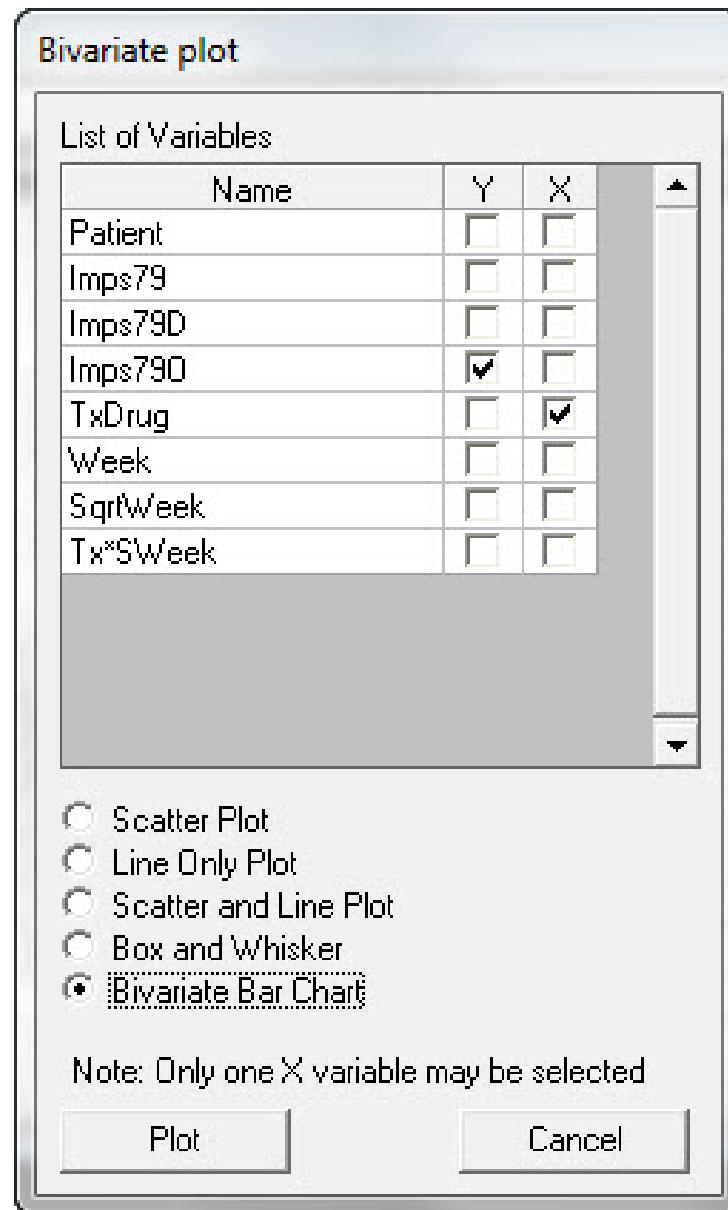
Select “File” > “Data-based Graphs” > “Bivariate”



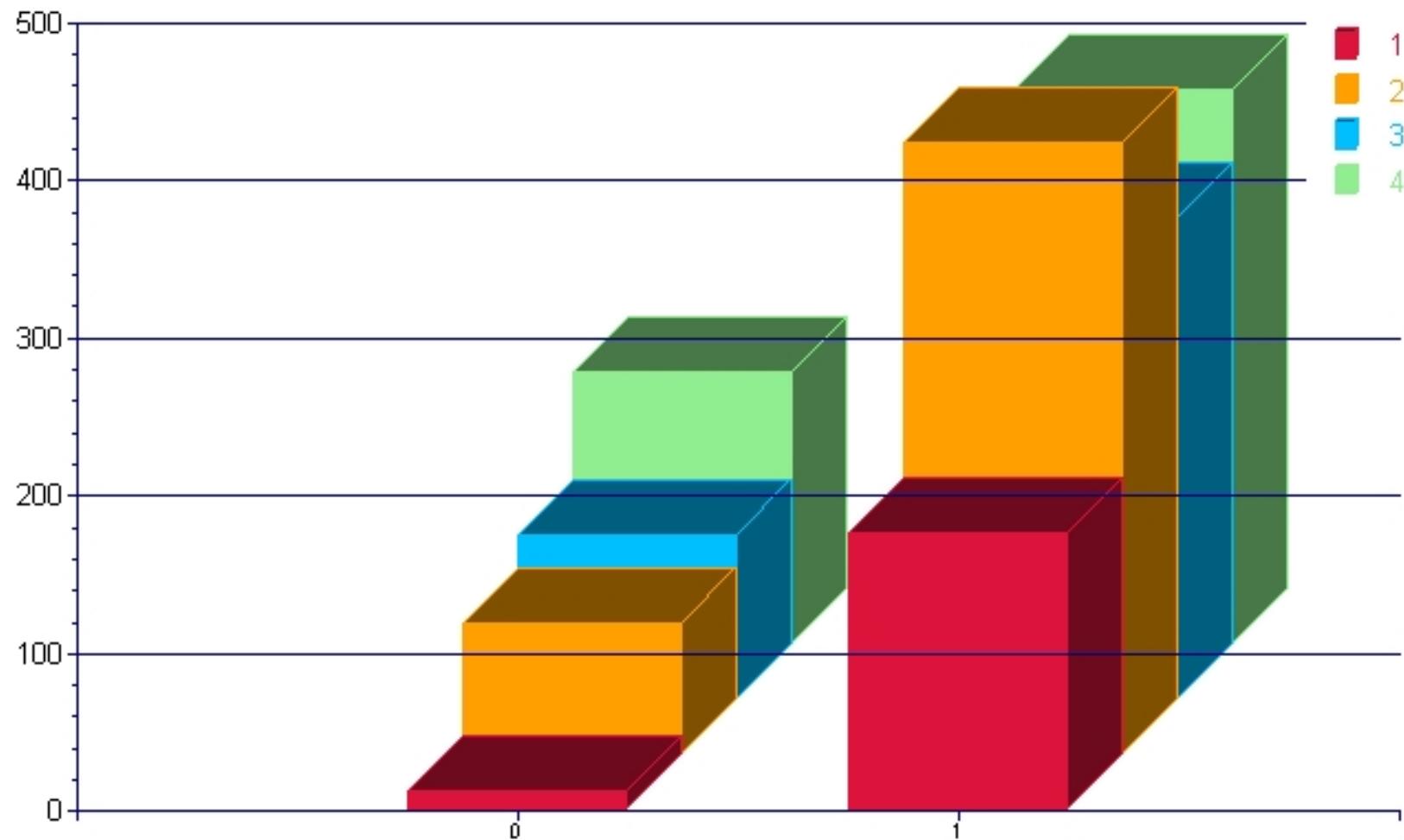
## Imps79O vs. Week



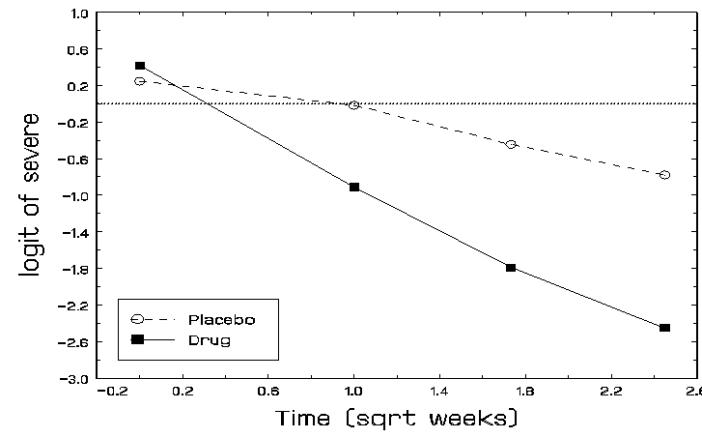
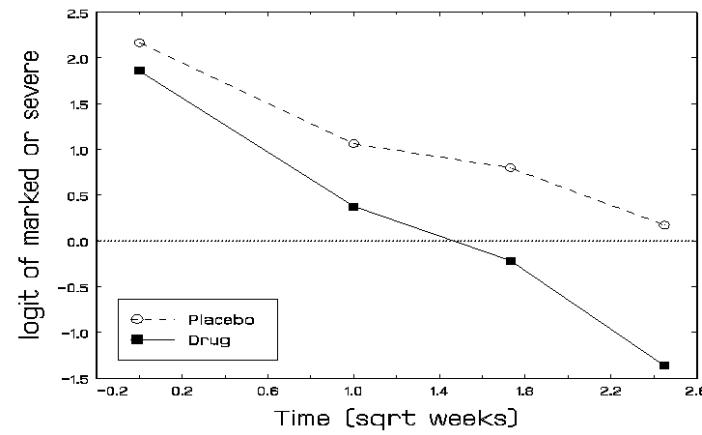
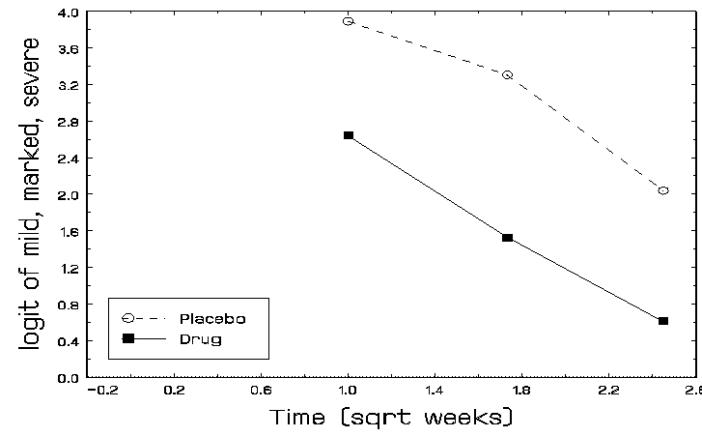
Select “File” > “Data-based Graphs” > “Bivariate”



## Imps79O vs. TxDrug



# Observed Logits across Time by Condition



## Within-Subjects / Between-Subjects components

Within-subjects model - level 1    ( $j = 1, \dots, n_i$  obs)

$$\lambda_{ijc} = \gamma_c - [b_{0i} + b_{1i}\sqrt{Week_j}]$$

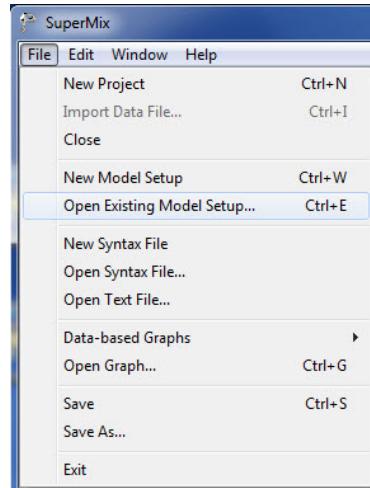
Between-subjects model - level 2    ( $i = 1, \dots, N$  subjects)

$$b_{0i} = \beta_0 + \beta_2 Grp_i + v_{0i}$$

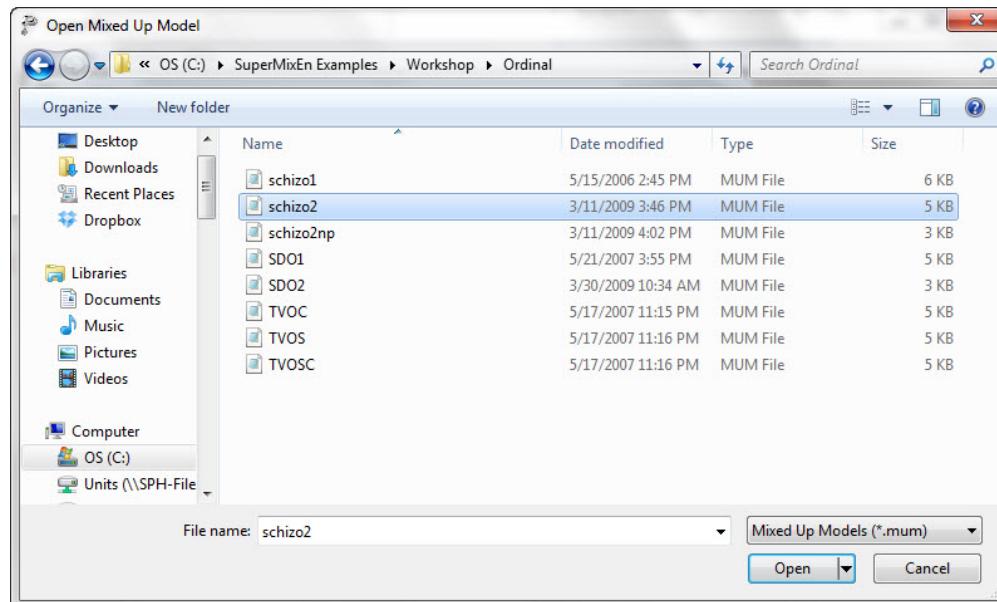
$$b_{1i} = \beta_1 + \beta_3 Grp_i + v_{1i}$$

$$\boldsymbol{v}_i \sim \mathcal{NID}(\mathbf{0}, \boldsymbol{\Sigma}_v)$$

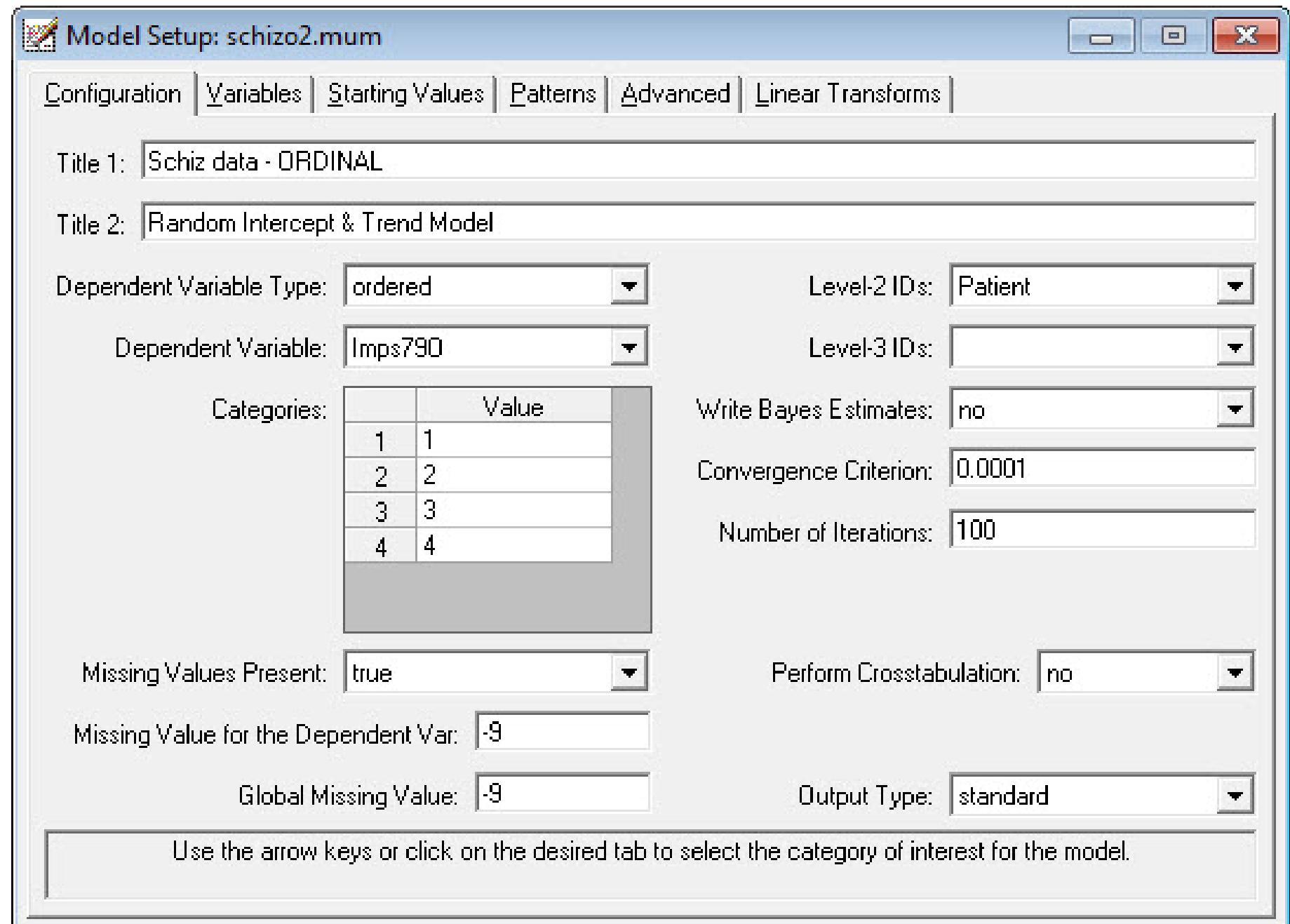
Under “File” click on “Open Existing Model Setup”



Open C:\SuperMixEn Examples\Workshop\Ordinal\schizo2.mum  
(or C:\SuperMixEn Student Examples\Workshop\Ordinal\schizo2.mum)



Note that “Dependent Variable Type” is “ordered”



Note the lack of TxDrug as an explanatory variable

Model Setup: schizo2.mum

Configuration Variables Starting Values Patterns Advanced Linear Transforms

Available	E	2
Patient	<input type="checkbox"/>	<input checked="" type="checkbox"/>
Imps79	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
Imps79D	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
Imps79O	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
TxDrag	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
Week	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
SqrWweek	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
Tx*SWweek	<input checked="" type="checkbox"/>	<input type="checkbox"/>

Explanatory Variables

SqrWweek  
Tx\*SWweek

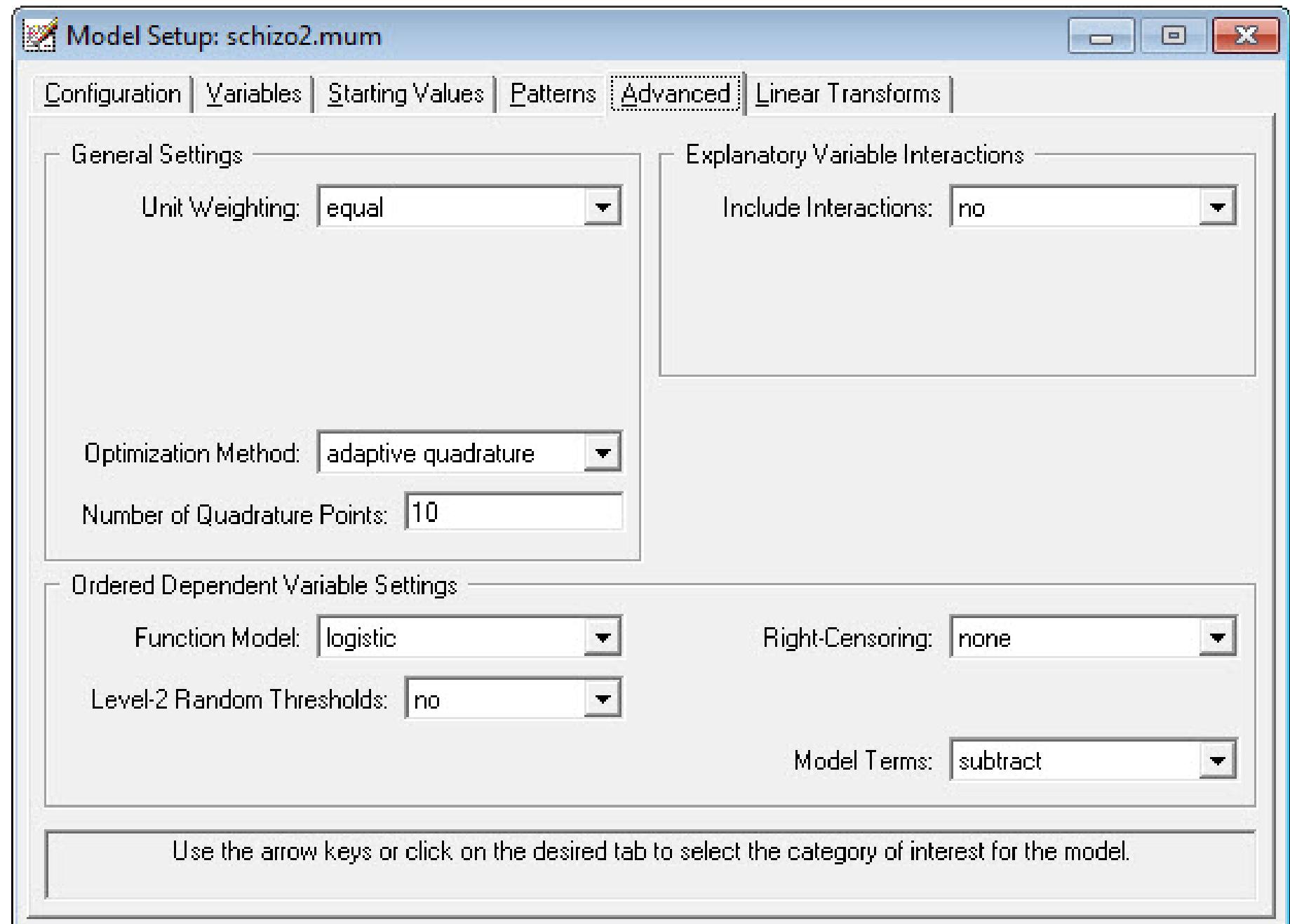
L-2 Random Effects

SqrWweek

Include Intercept

Select the columns of the spreadsheet to be used as explanatory variables and random effects.

Make sure “Optimization Method” is set to “adaptive quadrature”



## Note: Cumulative Logit link function

```
schizo2.out

o=====o
| Schiz data - ORDINAL          |
| Random Intercept & Trend Model |
o=====o

Model and Data Descriptions

Sampling Distribution = Multinomial
Link Function = Cumulative Logit
Number of Level-2 Units = 437
Number of Level-1 Units = 1603
Number of Level-1 Units per Level-2 Unit =
 4   4   3   4   4   4   4   4   4   3   4   4
 4   2   3   4   3   4   3   4   4   4   3   3
 2   4   4   4   4   4   3   4   4   4   4   4
 4   4   4   4   2   3   4   3   4   4   4   3
 4   4   2   2   4   5   4   2   4   4   3   4
 4   3   2   3   4   4   4   4   4   4   2   4
 4   4   5   4   4   2   2   4   2   4   4   3
 3   4   4   4   4   4   4   4   4   3   3   4
 2   3   4   4   4   2   5   3   4   4   2   4
 4   4   2   4   4   4   4   4   4   4   4   4
 5   2   4   3   4   4   2   2   4   4   4   4
 4   2   4   4   4   4   4   4   4   4   4   4
 4   4   4   2   4   4   2   4   4   4   3   4
 2   4   4   3   2   3   4   4   3   3   4   3
 4   4   4   4   4   4   4   4   4   4   4   4
 4   4   2   3   3   5   4   3   4   4   3   2
 4   4   4   4   4   3   3   4   4   4   4   4
 4   4   4   4   4   2   3   4   4   4   3   4
 4   4   4   4   2   3   4   3   4   4   2   4
```

Save As... Close

Category response indicators (IMPS79O1-IMPS79O4); results of fixed-effects model (to be ignored, or for comparison purposes)

Descriptive statistics for all the variables in the model				
Variable	Minimum	Maximum	Mean	Standard Deviation
Imps79O1	0.0000	1.0000	0.1185	0.3233
Imps79O2	0.0000	1.0000	0.2957	0.4565
Imps79O3	0.0000	1.0000	0.2570	0.4371
Imps79O4	0.0000	1.0000	0.3288	0.4699
SqrtWeek	0.0000	2.4495	1.2204	0.8965
Tx*SWeek	0.0000	2.4495	0.9442	0.9454

Results for the model without any random effects				
Save As...	Close			

## schizo2.out

```
o=====o
| Optimization Method: Adaptive Quadrature |
o=====o
```

Number of quadrature points = 10  
 Number of free parameters = 8  
 Number of iterations used = 5  
  
 -2lnL (deviance statistic) = 3325.51347  
 Akaike Information Criterion 3341.51347  
 Schwarz Criterion 3384.55053

## Estimated regression weights

Parameter	Estimate	Standard Error	z Value	P Value
Threshold1	-7.3662	0.3711	-19.8494	0.0000
Threshold2	-3.4647	0.2438	-14.2121	0.0000
Threshold3	-0.8577	0.1850	-4.6353	0.0000
SqrtWeek	-0.8996	0.1897	-4.7430	0.0000
Tx*SWeek	-1.6740	0.2081	-8.0456	0.0000

## Odds Ratio and 95% Odds Ratio Confidence Intervals

Parameter	Estimate	Odds Ratio	Lower	Upper
Threshold1	-7.3662	0.0006	0.0003	0.0013
Threshold2	-3.4647	0.0313	0.0194	0.0504
Threshold3	-0.8577	0.4242	0.2951	0.6096
SqrtWeek	-0.8996	0.4067	0.2805	0.5899
Tx*SWeek	-1.6740	0.1875	0.1247	0.2819

 Save As... Close

## Estimated level 2 variances and covariances

Parameter	Estimate	Standard Error	z Value	P Value
<hr/>				
intercept/intercept	7.0193	1.3188	5.3224	0.0000
SqrtWeek/intercept	-1.5174	0.5321	-2.8516	0.0043
SqrtWeek/SqrtWeek	2.0134	0.4178	4.8186	0.0000

## Level 2 covariance matrix

	intercept	SqrtWeek
intercept	7.019260	
SqrtWeek	-1.517353	2.013390

## Level 2 correlation matrix

	intercept	SqrtWeek
intercept	1.000000	
SqrtWeek	-0.403624	1.000000

o=====o  
| SuperMix used 1.11 seconds CPU |  
o=====o

Save As...

Close

## SCHIZo2.out

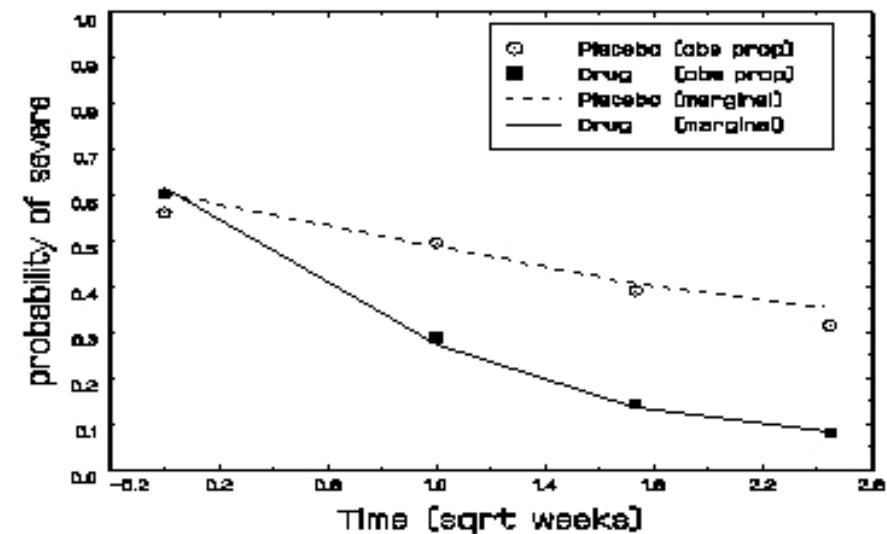
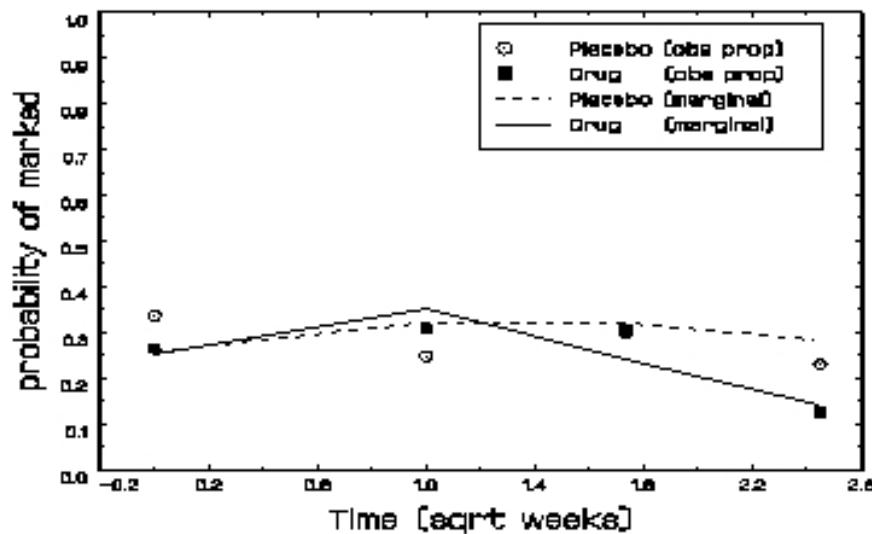
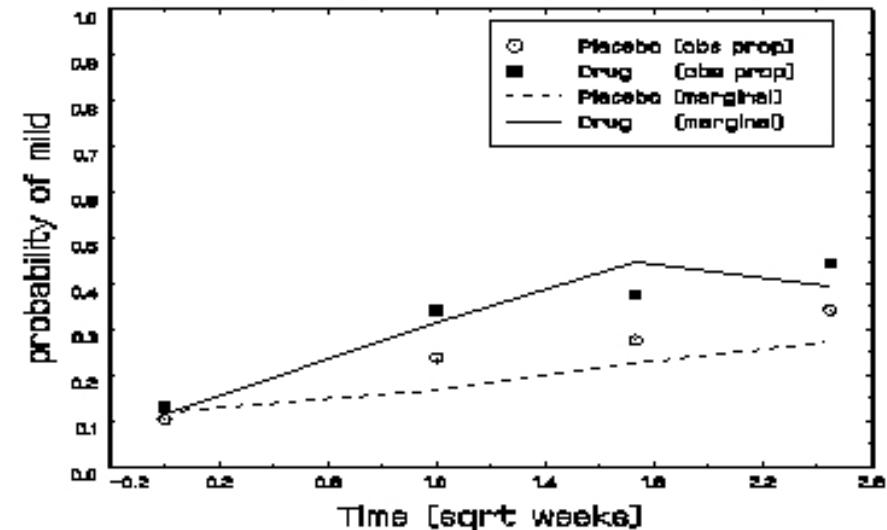
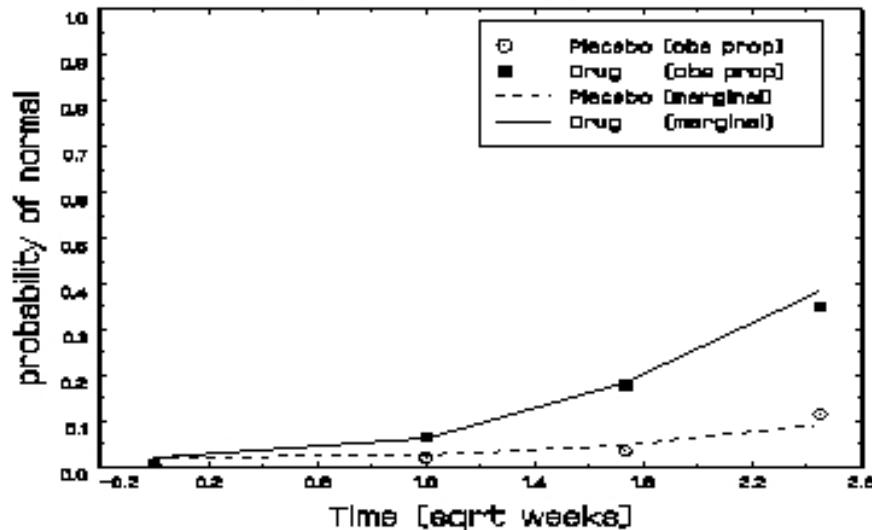
## Population Average Estimates

Parameter	Estimate	Standard Error	z Value	P Value
Threshold1	-4.5153	0.2582	-17.4851	0.0000
Threshold2	-1.7260	0.1720	-10.0338	0.0000
Threshold3	0.4765	0.1948	2.4459	0.0145
SqrtWeek	-0.8041	0.1469	-5.4753	0.0000
Tx*SWeek	-0.9018	0.1113	-8.1021	0.0000

## Odds Ratio and 95% Odds Ratio Confidence Intervals

Parameter	Estimate	Odds Ratio	Bounds	
			Lower	Upper
Threshold1	-4.5153	0.0109	0.0066	0.0181
Threshold2	-1.7260	0.1780	0.1270	0.2494
Threshold3	0.4765	1.6104	1.0993	2.3592
SqrtWeek	-0.8041	0.4475	0.3356	0.5967
Tx*SWeek	-0.9018	0.4058	0.3263	0.5048

# Model Fit of Observed Proportions



**SAS IML code:** SCHZOFIT.SAS - computing marginal probabilities - ordinal model  
adapted from syntax at <http://www.uic.edu/classes/bstt/bstt513/> (Week 12)

---

```
TITLE1 'NIMH Schizophrenia Data - Estimated Marginal Probabilities';
PROC IML;
/* Results from random intercept and trend model */;
/* using Population Average Estimates */;
x0 = { 0.00000 0,
        1.00000 0,
        1.73205 0,
        2.44949 0};
x1 = { 0.00000 0.00000,
        1.00000 1.00000,
        1.73205 1.73205,
        2.44949 2.44949};

beta    = { -.8041, -.9018};
thresh = { -4.5153, -1.726, .4765};
```

```

za0 = (thresh[1] - x0*beta) ;
zb0 = (thresh[2] - x0*beta) ;
zc0 = (thresh[3] - x0*beta) ;
za1 = (thresh[1] - x1*beta) ;
zb1 = (thresh[2] - x1*beta) ;
zc1 = (thresh[3] - x1*beta) ;

grp0a = 1 / ( 1 + EXP(- za0));
grp0b = 1 / ( 1 + EXP(- zb0));
grp0c = 1 / ( 1 + EXP(- zc0));
grp1a = 1 / ( 1 + EXP(- za1));
grp1b = 1 / ( 1 + EXP(- zb1));
grp1c = 1 / ( 1 + EXP(- zc1));

print 'Random intercept and trend model';
print using Population Average Estimates';
print 'marginal prob for group 0 - catg 1' grp0a [FORMAT=8.4];
print 'marginal prob for group 0 - catg 2' (grp0b-grp0a) [FORMAT=8.4];
print 'marginal prob for group 0 - catg 3' (grp0c-grp0b) [FORMAT=8.4];
print 'marginal prob for group 0 - catg 4' (1-grp0c) [FORMAT=8.4];
print 'marginal prob for group 1 - catg 1' grp1a [FORMAT=8.4];
print 'marginal prob for group 1 - catg 2' (grp1b-grp1a) [FORMAT=8.4];
print 'marginal prob for group 1 - catg 3' (grp1c-grp1b) [FORMAT=8.4];
print 'marginal prob for group 1 - catg 4' (1-grp1c) [FORMAT=8.4];

```

# Proportional and Non-proportional Odds

*Proportional Odds model*

$$\log \left[ \frac{P(Y_{ij} \leq c)}{1 - P(Y_{ij} \leq c)} \right] = \gamma_c - [\mathbf{x}'_{ij} \boldsymbol{\beta} + \mathbf{z}'_{ij} \boldsymbol{\nu}_i]$$

with  $\boldsymbol{\nu}_i \sim N(\mathbf{0}, \boldsymbol{\Sigma}_{\nu})$

- relationship between the explanatory variables and the cumulative logits does not depend on  $c$
- effects of  $\mathbf{x}$  variables DO NOT vary across the  $C - 1$  cumulative logits

## *Non-Proportional/Partial Proportional Odds model*

$$\log \left[ \frac{P(Y_{ij} \leq c)}{1 - P(Y_{ij} \leq c)} \right] = \gamma_{0c} - [\boldsymbol{u}'_{ij} \boldsymbol{\gamma}_c + \boldsymbol{x}'_{ij} \boldsymbol{\beta} + \boldsymbol{z}'_{ij} \boldsymbol{v}_i]$$

$\boldsymbol{u}_{ij}$  =  $h \times 1$  vector for the set of  $h$  covariates for which proportional odds is not assumed

- effects of  $\boldsymbol{u}$  variables DO vary across the  $C - 1$  cumulative logits
- more flexible model for ordinal response relations
- can be used to empirically test proportional odds assumption

**Proportional Odds Assumption:** covariate effects are the same across all cumulative logits

group	<i>Response</i>			total
	Absent	Mild	Severe	
control	27	46	27	100
cumulative odds	$\frac{27}{73} = .37$	$\frac{73}{27} = 2.7$		
<i>logit</i>	-1	1		
treatment	38	44	18	100
cumulative odds	$\frac{38}{62} = .61$	$\frac{82}{18} = 4.6$		
<i>logit</i>	-.5	1.5		

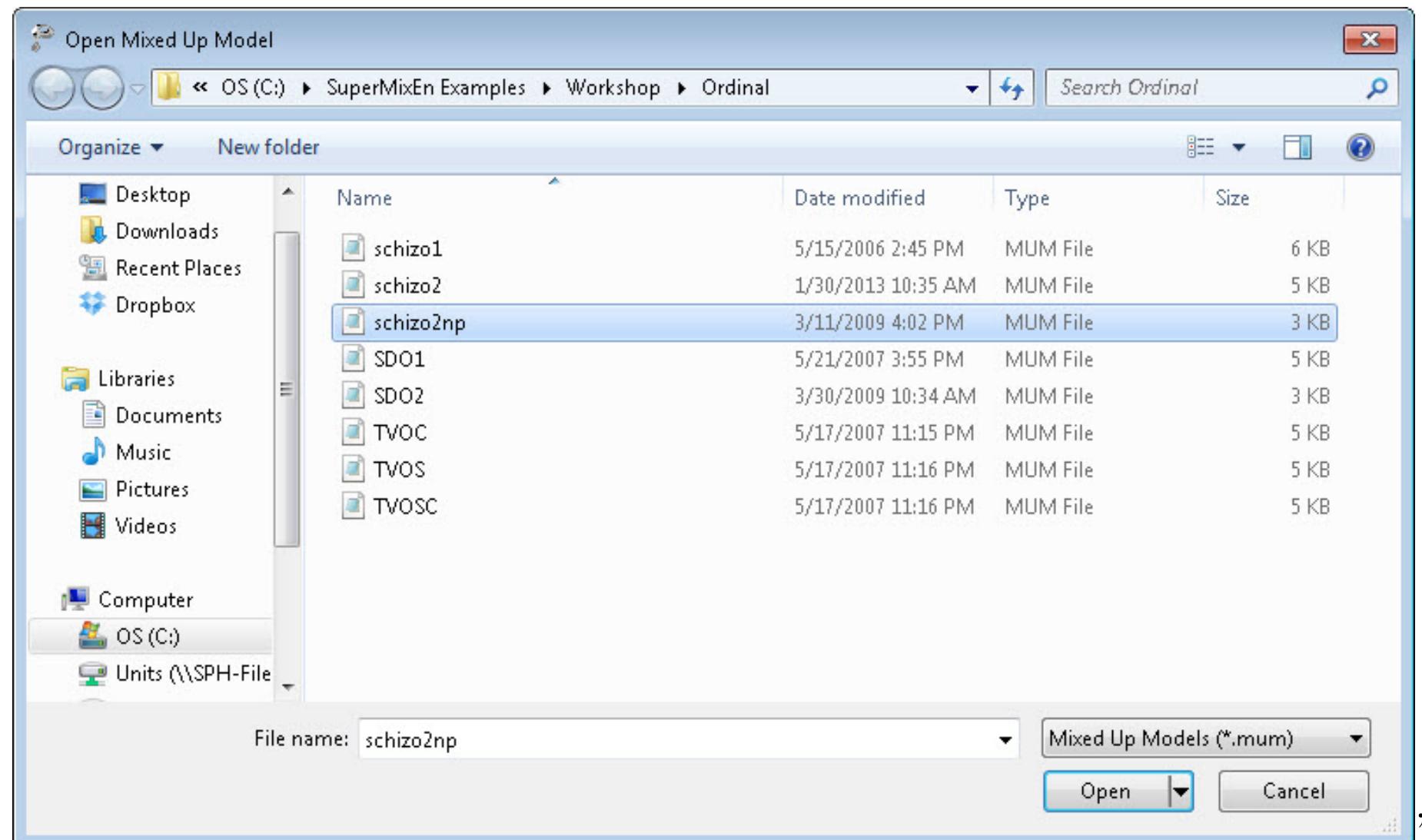
$\Rightarrow$  group difference = .5 for both cumulative logits

**Non-Proportional Odds:** covariate effects vary across the cumulative logits

group	<i>Response</i>			total
	Absent	Mild	Severe	
control	27	46	27	100
cumulative odds	$\frac{27}{73} = .37$	$\frac{73}{27} = 2.7$		
<i>logit</i>	-1	1		
treatment	28	60	12	100
cumulative odds	$\frac{28}{72} = .39$	$\frac{88}{12} = 7.3$		
<i>logit</i>	-.95	2		

$\Rightarrow$  *UNEQUAL group difference across cumulative logits*

Open C:\SuperMixEn Examples\Workshop\Ordinal\schizo2np.mum  
(or C:\SuperMixEn Student Examples\Workshop\Ordinal\schizo2np.mum)



Note that “Dependent Variable Type” is “ordered”

Model Setup: schizo2np.mum

Configuration | Variables | Starting Values | Patterns | Advanced | Linear Transforms

Title 1: Schiz data - ORDINAL - NON PROPORTIONAL ODDS

Title 2: Random Intercept & Trend Model

Dependent Variable Type: ordered

Level-2 IDs: Patient

Dependent Variable: Imps790

Level-3 IDs:

Categories:

	Value
1	1
2	2
3	3
4	4

Write Bayes Estimates: no

Convergence Criterion: 0.0001

Number of Iterations: 100

Missing Values Present: true

Perform Crosstabulation: no

Missing Value for the Dependent Var: -9

Global Missing Value: -9

Output Type: standard

Use the arrow keys or click on the desired tab to select the category of interest for the model.

Two explanatory variables: `SqrtWeek` and `Tx*SWeek`

Model Setup: schizo2np.mum

Configuration Variables Starting Values Patterns Advanced Linear Transforms

Available	E	2
Patient	<input type="checkbox"/>	<input type="checkbox"/>
Imps79	<input type="checkbox"/>	<input type="checkbox"/>
Imps79D	<input type="checkbox"/>	<input type="checkbox"/>
Imps79O	<input type="checkbox"/>	<input type="checkbox"/>
TxDrug	<input type="checkbox"/>	<input type="checkbox"/>
Week	<input type="checkbox"/>	<input type="checkbox"/>
<b>SqrtWeek</b>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
<b>Tx*SWeek</b>	<input checked="" type="checkbox"/>	<input type="checkbox"/>

Explanatory Variables

`SqrtWeek`  
`Tx*SWeek`

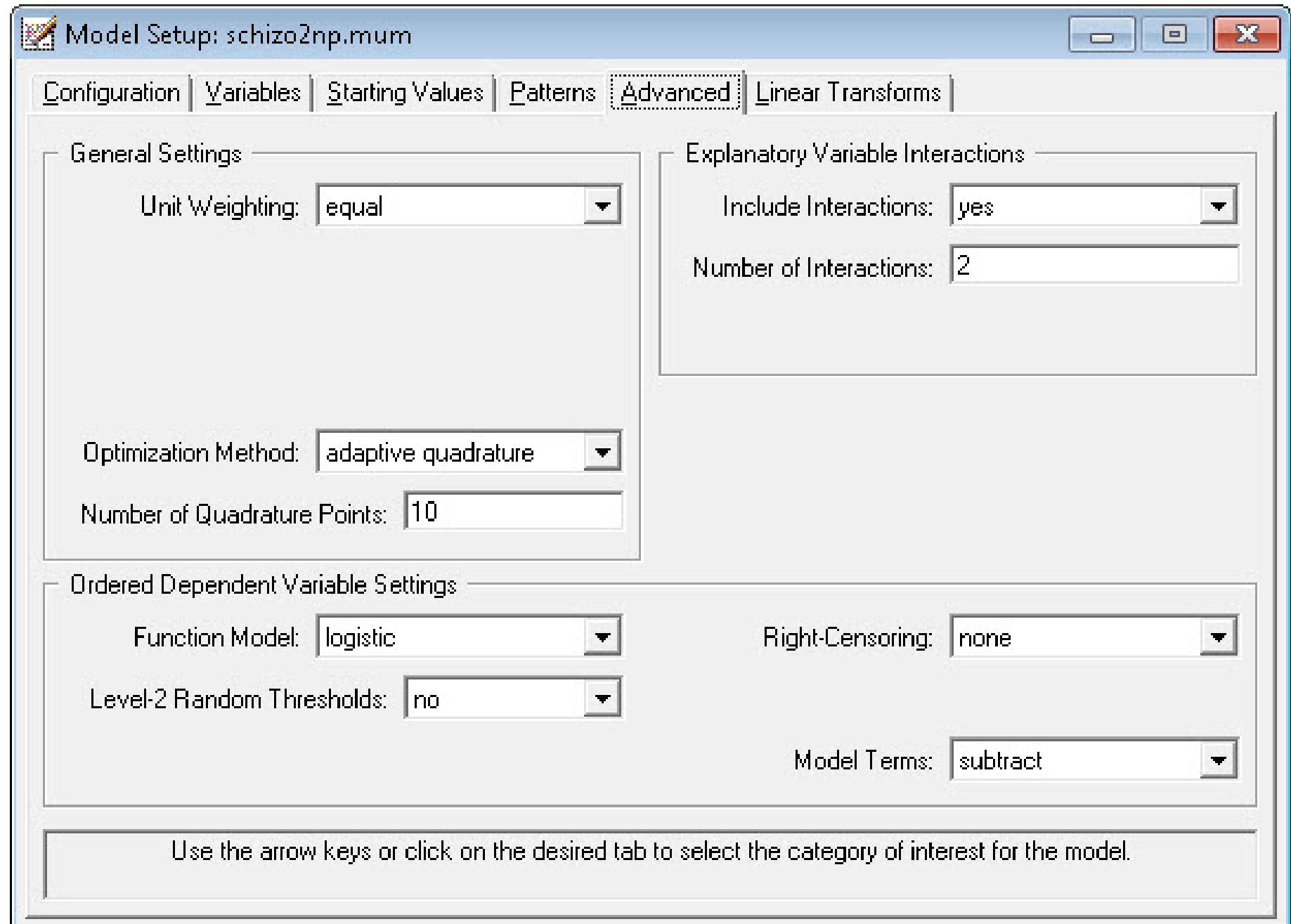
L-2 Random Effects

`SqrtWeek`

Include Intercept

Select the columns of the spreadsheet to be used as explanatory variables and random effects.

“Explanatory Variable Interactions” - both are selected



schizo2np.out

```

=====
| Optimization Method: Adaptive Quadrature |
=====

Number of quadrature points =          10
Number of free parameters =          12
Number of iterations used =          6

-2lnL (deviance statistic) =      3324.16153
Akaike Information Criterion      3348.16153
Schwarz Criterion                3412.71711

Estimated regression weights

      Standard
Parameter   Estimate    Error   z Value   P Value
-----
Threshold1  -7.4687    0.4860  -15.3681  0.0000
Threshold2  -3.5799    0.2779  -12.8808  0.0000
Threshold3  -0.8100    0.1935  -4.1848   0.0000
SqrtWeek    -0.9506    0.3307  -2.8741   0.0041
Tx*SWeek    -1.6827    0.2879  -5.8456   0.0000

Interactions of predictors with: Threshold2
SqrtWeek     -0.0826    0.2991  -0.2761   0.7825
Tx*SWeek     0.0799    0.2496  0.3199   0.7490

Interactions of predictors with: Threshold3
SqrtWeek     0.1171    0.3332  0.3515   0.7252
Tx*SWeek    -0.0004    0.2870  -0.0013   0.9990

```

Save As...    Close

Proportional Odds Assumption Accepted:  $\chi^2_4 = 3325.51 - 3324.16 = 1.35$

# Linear Transforms

Fixed part of model:

$$\lambda_c = \hat{\gamma}_{0c} - [\hat{\beta}_1 \text{SqrtWeek} + \hat{\beta}_2 \text{Tx*SWeek} \\ + \hat{\gamma}_{1c} \text{SqrtWeek} + \hat{\gamma}_{2c} \text{Tx*SWeek}]$$

cumulative logit

variable	1 vs 2,3,4	1,2 vs 3,4	1,2,3, vs 4
SqrtWeek	$\hat{\beta}_1$	$\hat{\beta}_1 + \hat{\gamma}_{12}$	$\hat{\beta}_1 + \hat{\gamma}_{13}$
Tx*SWeek	$\hat{\beta}_2$	$\hat{\beta}_2 + \hat{\gamma}_{22}$	$\hat{\beta}_2 + \hat{\gamma}_{23}$

$H_0 : \beta_1 + \gamma_{12} = 0$ ; SqrtWeek effect is 0 on the 2nd cumulative logit

$$z = \frac{\hat{\beta}_1 + \hat{\gamma}_{12}}{SE(\hat{\beta}_1 + \hat{\gamma}_{12})}$$

**Model Setup: schizo2np.mum**

- Linear Transforms:**
  - SqrWeek Thresh2
  - Tx\*SWeek Thresh2
  - SqrWeek Thresh3
  - Tx\*SWeek Thresh3
- Add Transform**
- Copy Transform**
- Remove Transform**

	Value
SqrWeek	1
Tx*SWeek	

	Value
intercept variance	
intercept, SqrWeek	
SqrWeek variance	

	Value
1	
2	
3	

	Thresh 2	Thresh 3
SqrWeek	1	
Tx*SWeek		

Select the linear transform to review and edit its components.  
Type to change the transform's name in place.

**Model Setup: schizo2np.mum**

- Linear Transforms:**
  - SqrWeek Thresh2
  - Tx\*SWeek Thresh2
  - SqrWeek Thresh3
  - Tx\*SWeek Thresh3
- Add Transform**
- Copy Transform**
- Remove Transform**

	Value
SqrWeek	
Tx*SWeek	1

	Value
intercept variance	
intercept, SqrWeek	
SqrWeek variance	

	Value
1	
2	
3	

	Thresh 2	Thresh 3
SqrWeek		
Tx*SWeek		1

Select the linear transform to review and edit its components.  
Type to change the transform's name in place.

**Model Setup: schizo2np.mum**

- Linear Transforms:**
  - SqrWeek Thresh2
  - Tx\*SWeek Thresh2
  - SqrWeek Thresh3
  - Tx\*SWeek Thresh3
- Add Transform**
- Copy Transform**
- Remove Transform**

	Value
SqrWeek	1
Tx*SWeek	

	Value
intercept variance	
intercept, SqrWeek	
SqrWeek variance	

	Value
1	
2	
3	

	Thresh 2	Thresh 3
SqrWeek		1
Tx*SWeek		

Select the linear transform to review and edit its components.  
Type to change the transform's name in place.

**Model Setup: schizo2np.mum**

- Linear Transforms:**
  - SqrWeek Thresh2
  - Tx\*SWeek Thresh2
  - SqrWeek Thresh3
  - Tx\*SWeek Thresh3
- Add Transform**
- Copy Transform**
- Remove Transform**

	Value
SqrWeek	
Tx*SWeek	1

	Value
intercept variance	
intercept, SqrWeek	
SqrWeek variance	

	Value
1	
2	
3	

	Thresh 2	Thresh 3
SqrWeek		
Tx*SWeek		1

Select the linear transform to review and edit its components.  
Type to change the transform's name in place.

schizo2np.out

---

TESTING OF TRANSFORMS  
(General Linear Hypothesis Testing)

---

Coefficients		Estimate	Transform No.		
			1	2	3
1 Threshold1		-7.46875	0.0000	0.0000	0.0000
2 Threshold2		-3.57991	0.0000	0.0000	0.0000
3 Threshold3		-0.80996	0.0000	0.0000	0.0000
4 SqrtWeek		-0.95058	1.0000	0.0000	1.0000
5 Tx*SWeek		-1.68268	0.0000	1.0000	0.0000
6 SqrtWeek	*Threshold2	-0.08259	1.0000	0.0000	0.0000
7 Tx*SWeek	*Threshold2	0.07985	0.0000	1.0000	0.0000
8 SqrtWeek	*Threshold3	0.11712	0.0000	0.0000	1.0000
9 Tx*SWeek	*Threshold3	-0.00037	0.0000	0.0000	0.0000
10 Var(intercept)		7.37046	0.0000	0.0000	0.0000
11 Cov(SqrtWeek,intercept)		-1.67047	0.0000	0.0000	0.0000
12 Var(SqrtWeek)		2.03449	0.0000	0.0000	0.0000
Transform Estimate			-1.0332	-1.6028	-0.8335
Standard Error			0.2240	0.2287	0.2063
Z-Statistic			-4.6133	-7.0090	-4.0392
Exceedence Pobability			0.0000	0.0000	0.0001

Save As... Close

schizo2np.out

---

LINEAR TRANSFORMS (continued)

---

Coefficients	Estimate	Transform No.
		4
1 Threshold1	-7.46875	0.0000
2 Threshold2	-3.57991	0.0000
3 Threshold3	-0.80996	0.0000
4 SqrtWeek	-0.95058	0.0000
5 Tx*SWeek	-1.68268	1.0000
6 SqrtWeek *Threshold2	-0.08259	0.0000
7 Tx*SWeek *Threshold2	0.07985	0.0000
8 SqrtWeek *Threshold3	0.11712	0.0000
9 Tx*SWeek *Threshold3	-0.00037	1.0000
10 Var(intercept)	7.37046	0.0000
11 Cov(SqrtWeek,intercept)	-1.67047	0.0000
12 Var(SqrtWeek)	2.03449	0.0000
Transform Estimate		-1.6831
Standard Error		0.2370
Z-Statistic		-7.1012
Exceedence Pobability		0.0000

---

<   >

NIMH Schiz Study: Severity of Illness ( $N = 437$ )  
 Ordinal LR Estimates (se) - *random intercept and trend model*

	Proportional Odds Model	Non-Proportional Odds Model		
		1 vs 2,3,4	1,2 vs 3,4	1,2,3 vs 4
Time (sqrt week)	-0.900 (0.190)	-0.951 (0.331)	-1.033 (0.224)	-0.834 (0.206)
Drug by Time	-1.674 (0.208)	-1.683 (0.288)	-1.603 (0.229)	-1.683 (0.237)
$-2 \log L$	3325.51		3324.16	

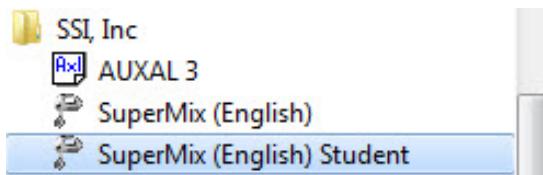
- Proportional Odds accepted ( $\chi^2_4 = 3325.51 - 3324.16 = 1.35$ )

## San Diego Homeless Research Project (Hough)

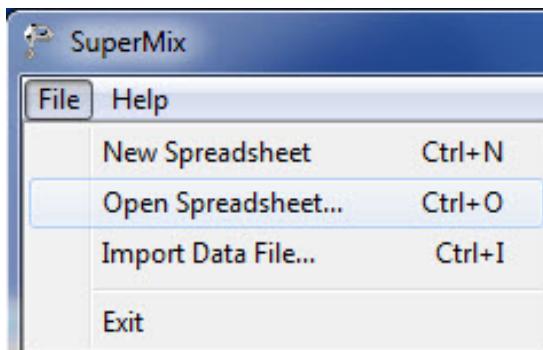
- 361 mentally ill subjects who were homeless or at very high risk of becoming homeless
- 2 conditions: HUD Section 8 rental certificates (yes/no)
- baseline and 6, 12, and 24 month follow-ups
- Categorical outcome: housing status
  - streets / shelters ( $Y = 0$ )
  - community / institutions ( $Y = 1$ )
  - independent ( $Y = 2$ )

*Question:* Do Section 8 certificates influence housing status across time?

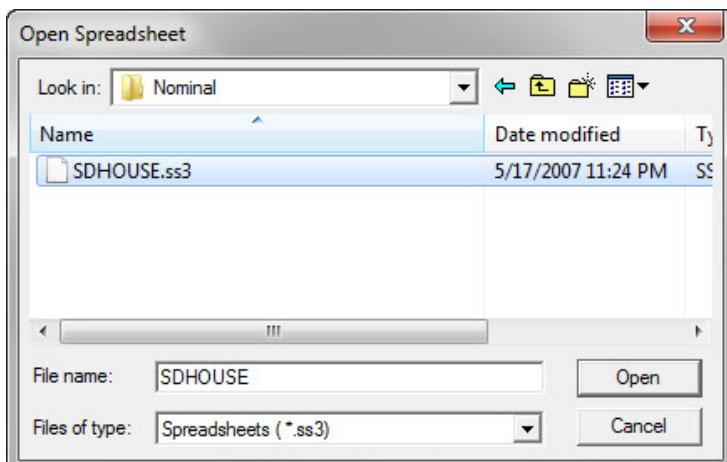
- Under SSI, Inc > “SuperMix (English)” or “SuperMix (English) Student”



- Under “File” click on “Open Spreadsheet”



- Open C:\SuperMixEn Examples\Workshop\Nominal\SDHOUSE.ss3  
(or C:\SuperMixEn Student Examples\Workshop\Nominal\SDHOUSE.ss3)



C:\SuperMixEn Examples\Workshop\Nominal\SDHOUSE.ss3

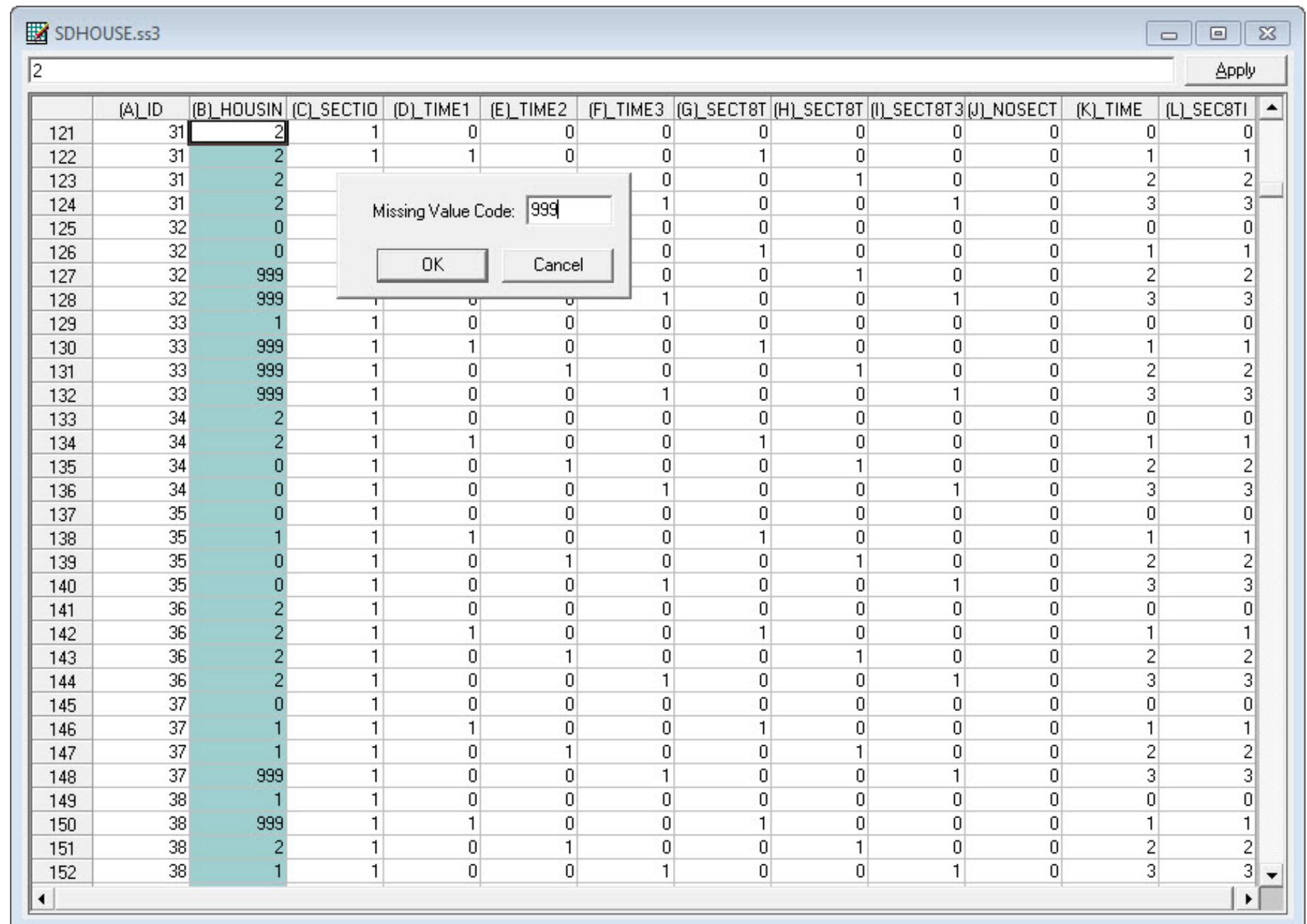
SDHOUSE.ss3

1

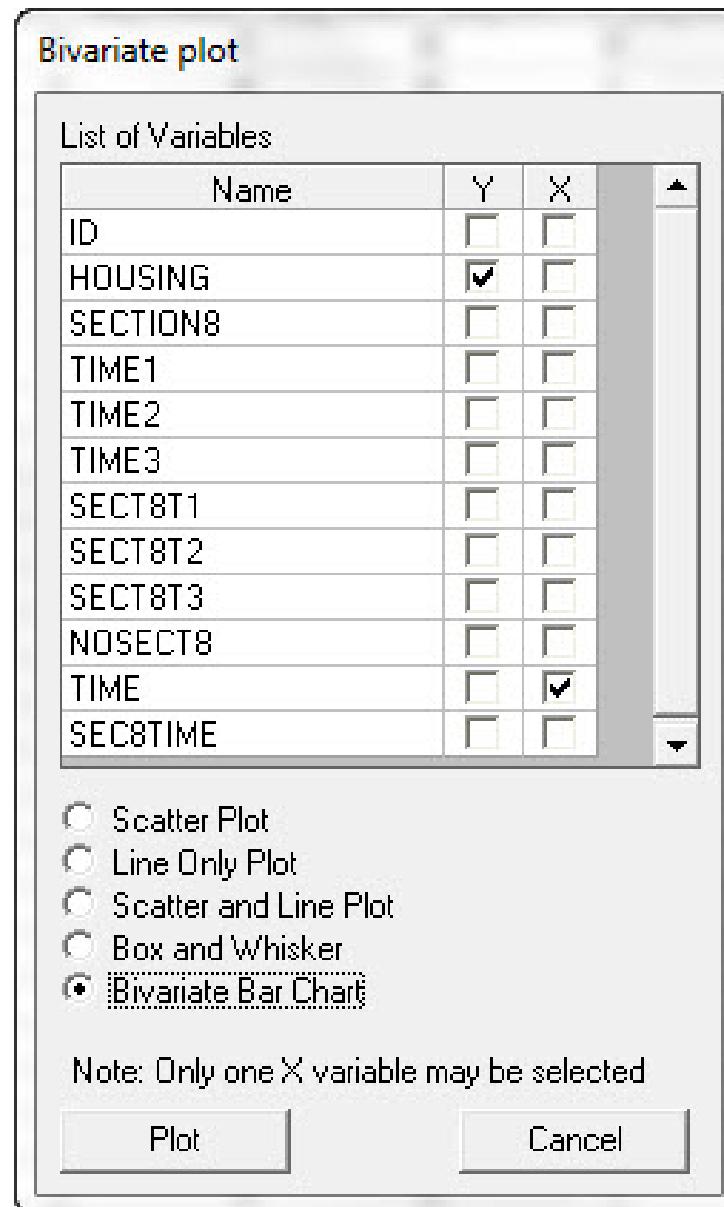
Apply

	(A)_ID	(B)_HOUSIN	(C)_SECTIO	(D)_TIME1	(E)_TIME2	(F)_TIME3	(G)_SECT8T	(H)_SECT8T	(I)_SECT8T3	(J)_NOSECT	(K)_TIME	(L)_SEC8TI
1	1	1	1	0	0	0	0	0	0	0	0	0
2	1	2	1	1	0	0	1	0	0	0	1	1
3	1	2	1	0	1	0	0	1	0	0	2	2
4	1	2	1	0	0	1	0	0	1	0	3	3
5	2	1	1	0	0	0	0	0	0	0	0	0
6	2	2	1	1	0	0	1	0	0	0	1	1
7	2	2	1	0	1	0	0	1	0	0	2	2
8	2	1	1	0	0	1	0	0	1	0	3	3
9	3	0	1	0	0	0	0	0	0	0	0	0
10	3	2	1	1	0	0	1	0	0	0	1	1
11	3	2	1	0	1	0	0	1	0	0	2	2
12	3	2	1	0	0	1	0	0	1	0	3	3
13	4	1	1	0	0	0	0	0	0	0	0	0
14	4	1	1	1	0	0	1	0	0	0	1	1
15	4	1	1	0	1	0	0	1	0	0	2	2
16	4	1	1	0	0	1	0	0	1	0	3	3
17	5	0	1	0	0	0	0	0	0	0	0	0
18	5	1	1	1	0	0	1	0	0	0	1	1
19	5	2	1	0	1	0	0	1	0	0	2	2
20	5	2	1	0	0	1	0	0	1	0	3	3
21	6	2	1	0	0	0	0	1	0	0	0	0
22	6	2	1	1	0	0	1	0	0	0	1	1
23	6	2	1	0	1	0	0	1	0	0	2	2
24	6	2	1	0	0	1	0	0	1	0	3	3
25	7	2	1	0	0	0	0	0	0	0	0	0
26	7	2	1	1	0	0	1	0	0	0	1	1
27	7	2	1	0	1	0	0	1	0	0	2	2
28	7	2	1	0	0	1	0	0	1	0	3	3
29	8	2	1	0	0	0	0	0	0	0	0	0
30	8	0	1	1	0	0	1	0	0	0	1	1
31	8	0	1	0	1	0	0	1	0	0	2	2
32	8	2	1	0	0	1	0	0	1	0	3	3

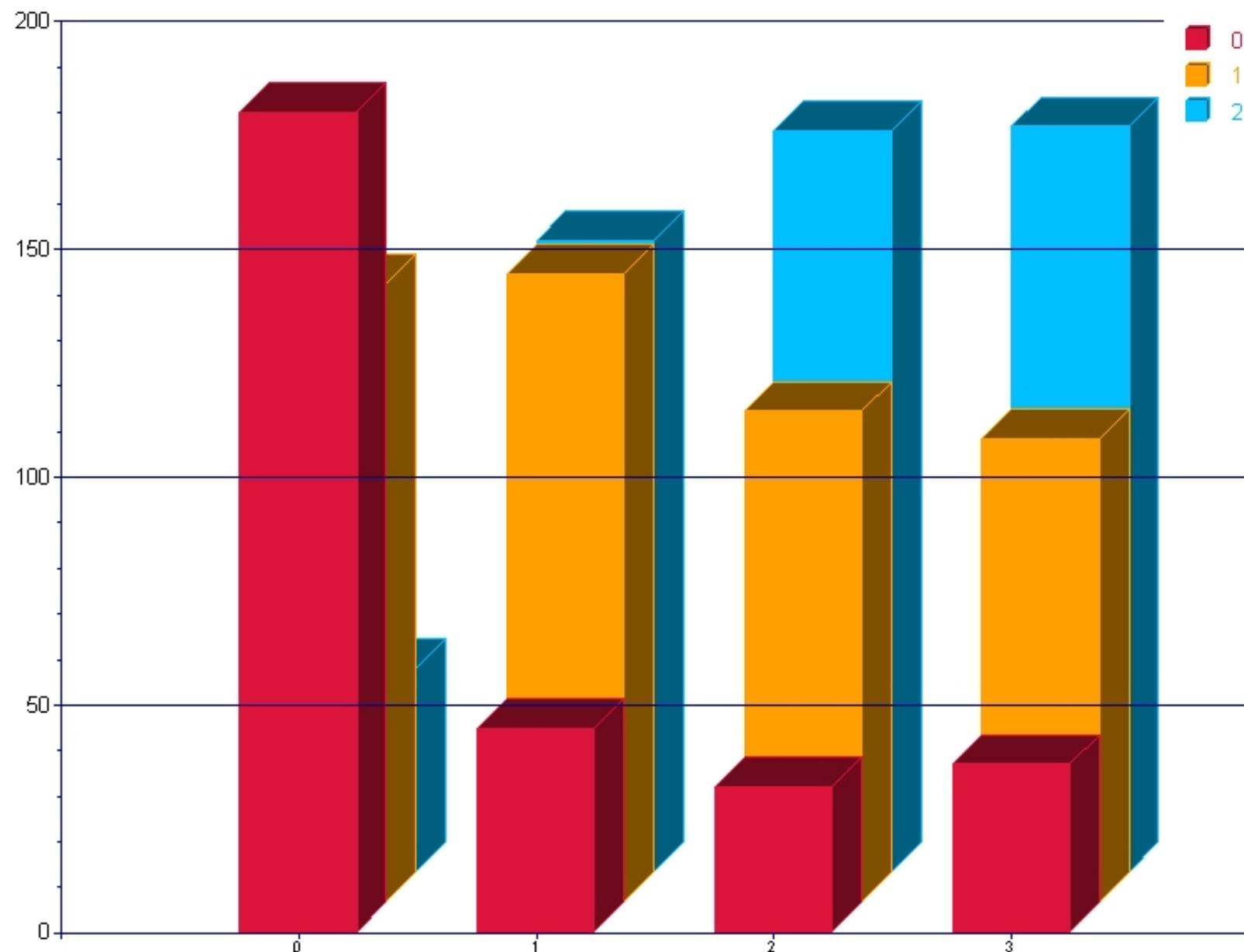
Select Housing column, then “Edit” > “Set Missing Value”



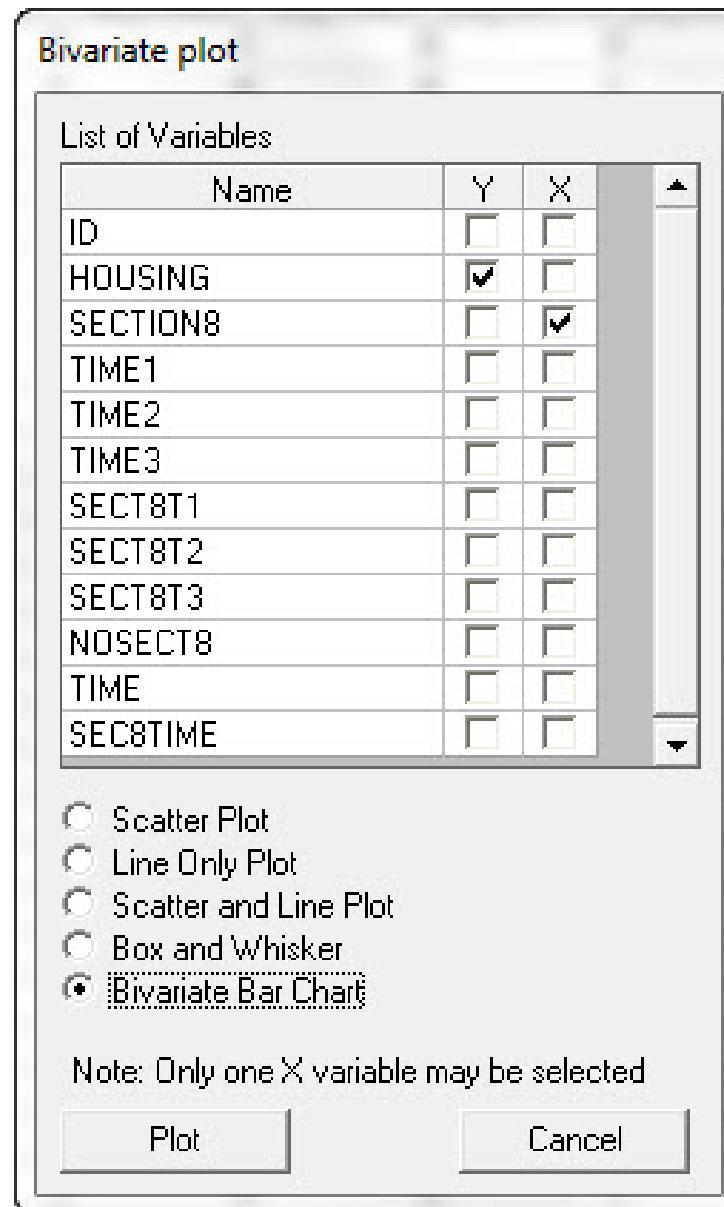
Select “File” > “Data-based Graphs” > “Bivariate”



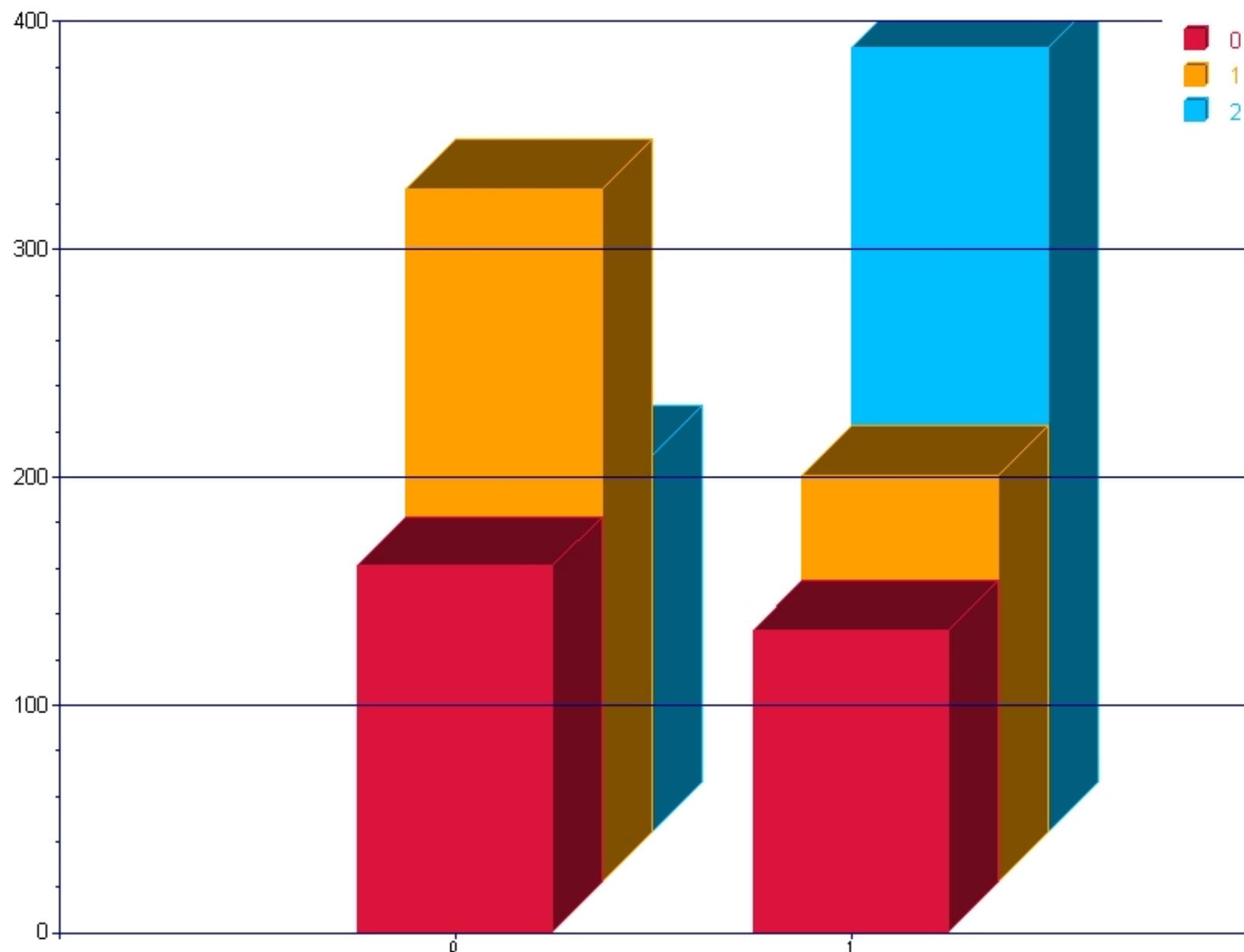
## HOUSING vs. TIME



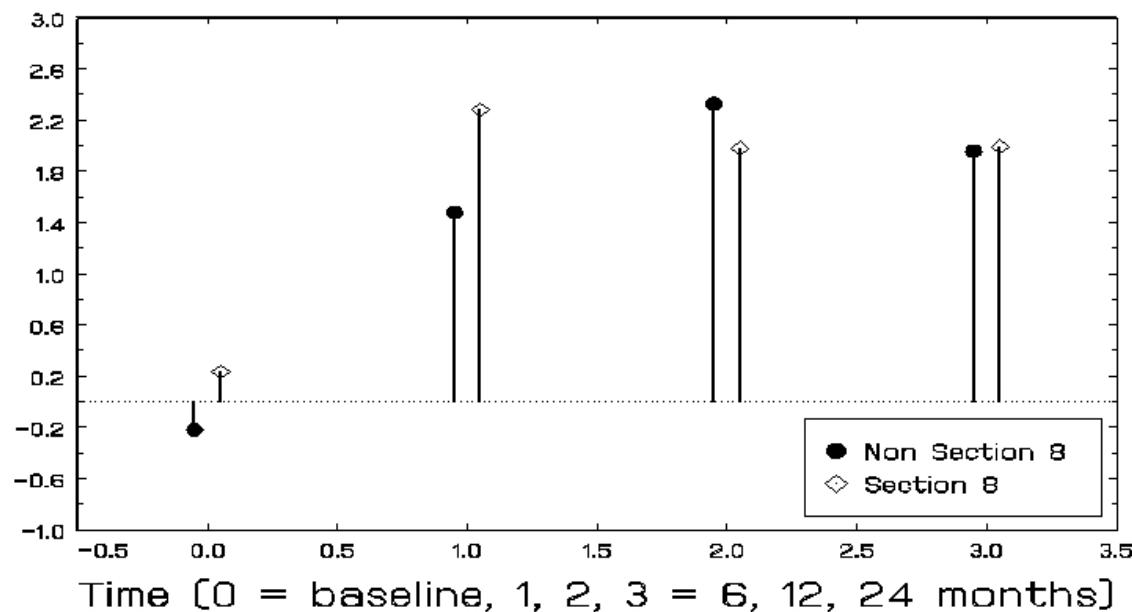
Select “File” > “Data-based Graphs” > “Bivariate”



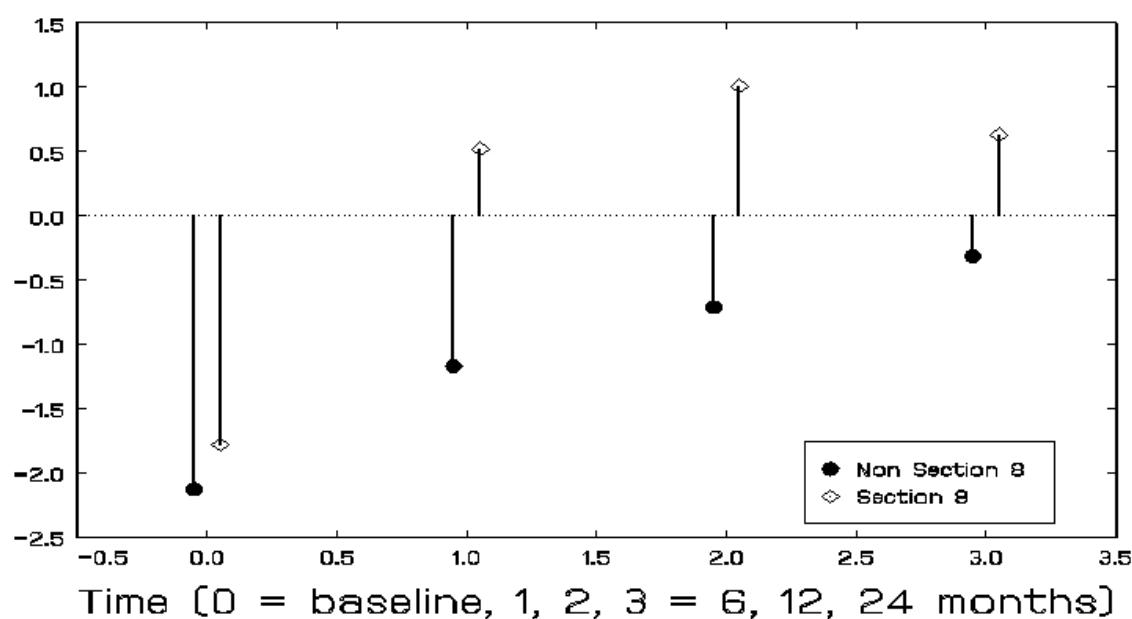
## HOUSING vs. SECTION8



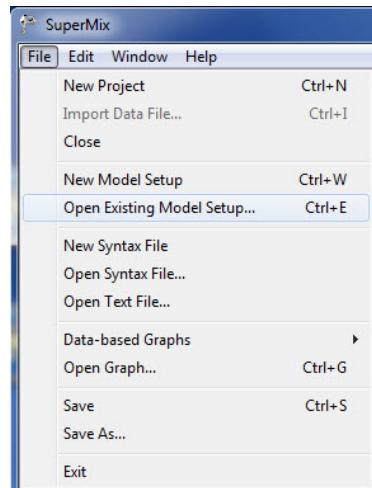
Empirical Logits - Ind & Comm vs Street



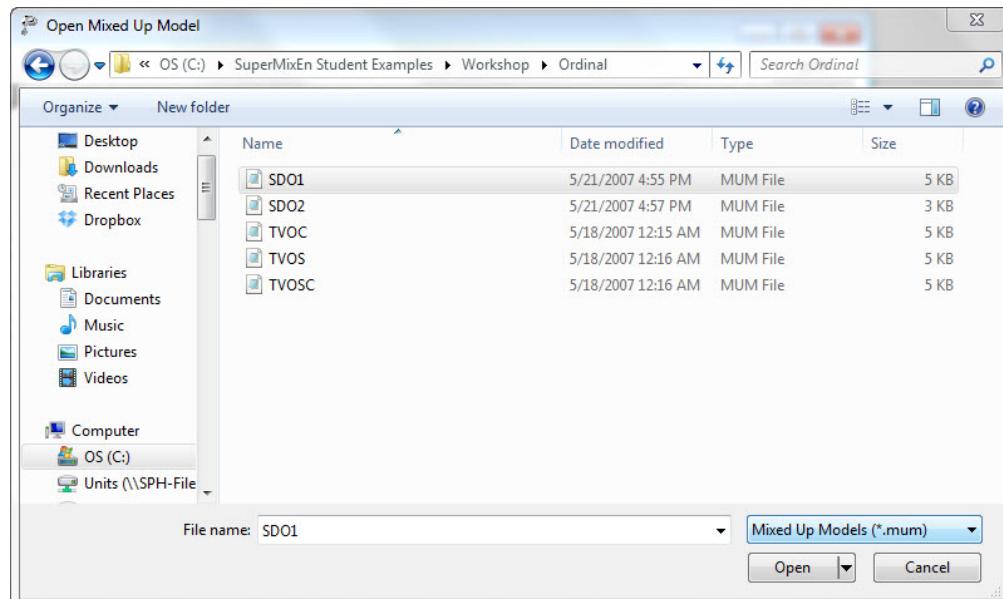
Empirical Logits - Ind vs Comm & Street



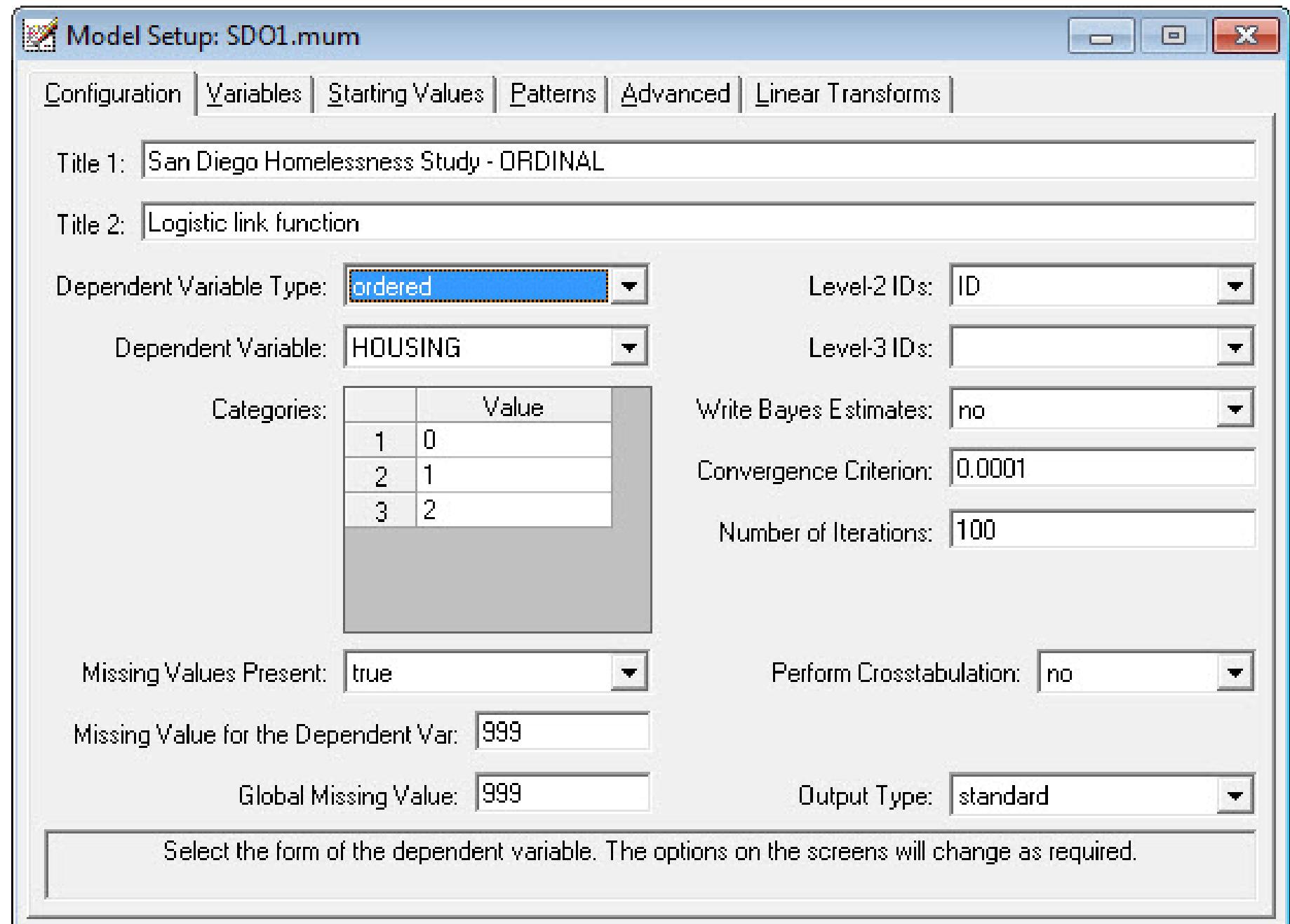
Under “File” click on “Open Existing Model Setup”



Open C:\SuperMixEn Examples\Workshop\Ordinal\SDO1.mum  
(or C:\SuperMixEn Student Examples\Workshop\Ordinal\SDO1.mum)



Note that “Dependent Variable Type” is “ordered”



All explanatory variables are indicator (dummy) variables

Model Setup: SDO1.mum

Configuration Variables Starting Values Patterns Advanced Linear Transforms

Available	E	2
ID	<input type="checkbox"/>	<input checked="" type="checkbox"/>
HOUSING	<input checked="" type="checkbox"/>	<input type="checkbox"/>
SECTION8	<input checked="" type="checkbox"/>	<input type="checkbox"/>
TIME1	<input checked="" type="checkbox"/>	<input type="checkbox"/>
TIME2	<input checked="" type="checkbox"/>	<input type="checkbox"/>
TIME3	<input checked="" type="checkbox"/>	<input type="checkbox"/>
SECT8T1	<input checked="" type="checkbox"/>	<input type="checkbox"/>
SECT8T2	<input checked="" type="checkbox"/>	<input type="checkbox"/>
SECT8T3	<input checked="" type="checkbox"/>	<input type="checkbox"/>
NOSECT8	<input type="checkbox"/>	<input type="checkbox"/>
TIME	<input type="checkbox"/>	<input type="checkbox"/>
SEC8TIME	<input type="checkbox"/>	<input type="checkbox"/>

Explanatory Variables

SECTION8  
TIME1  
TIME2  
TIME3  
SECT8T1  
SECT8T2  
SECT8T3

L-2 Random Effects

Include Intercept

Use the arrow keys or click on the desired tab to select the category of interest for the model.

Housing status across time: 1289 observations within 361 subjects  
 Ordinal Mixed Regression Model estimates and standard errors (se)

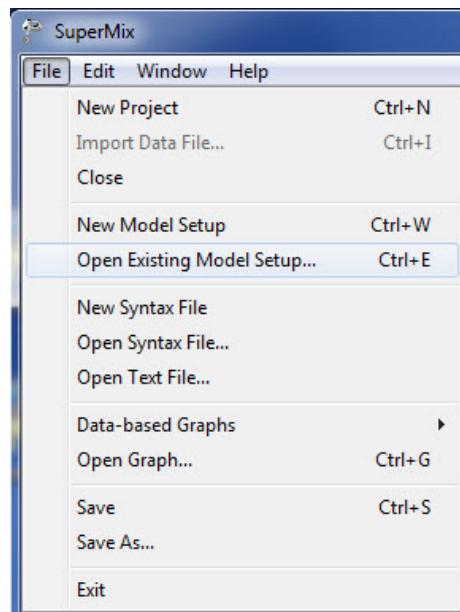
term	Proportional Odds		Non-Proportional Odds					
	estimate	se	estimate	se	estimate	se	difference	se
threshold <sub>1</sub>	.220	.198	.322	.207				
threshold <sub>2</sub>	<b>2.966</b>	.230			<b>2.700</b>	.298		
t1 (6 month)	<b>1.736</b>	.235	<b>2.298</b>	.303	<b>1.079</b>	.343	<b>-1.219</b>	.408
t2 (12 month)	<b>2.316</b>	.247	<b>3.346</b>	.387	<b>1.645</b>	.340	<b>-1.701</b>	.467
t3 (24 month)	<b>2.500</b>	.253	<b>2.822</b>	.348	<b>2.145</b>	.337	-.676	.422
section 8 (yes=1)	.497	.277	<b>.592</b>	.294	.323	.394	-.269	.384
section 8 by t1	<b>1.409</b>	.341	.566	.467	<b>2.024</b>	.471	<b>1.457</b>	.581
section 8 by t2	<b>1.173</b>	.354	-.958	.506	<b>2.017</b>	.476	<b>2.975</b>	.600
section 8 by t3	.638	.349	-.366	.480	<b>1.073</b>	.464	<b>1.440</b>	.573
subject var	<b>2.134</b>	.354	<b>2.128</b>	.353	<i>(ICC ≈ .4)</i>			
−2 log <i>L</i>	2274.39		2222.25		$(\chi^2_7 = 52.14)$			

**bold** indicates  $p < .05$       *italic* indicates  $.05 < p < .10$

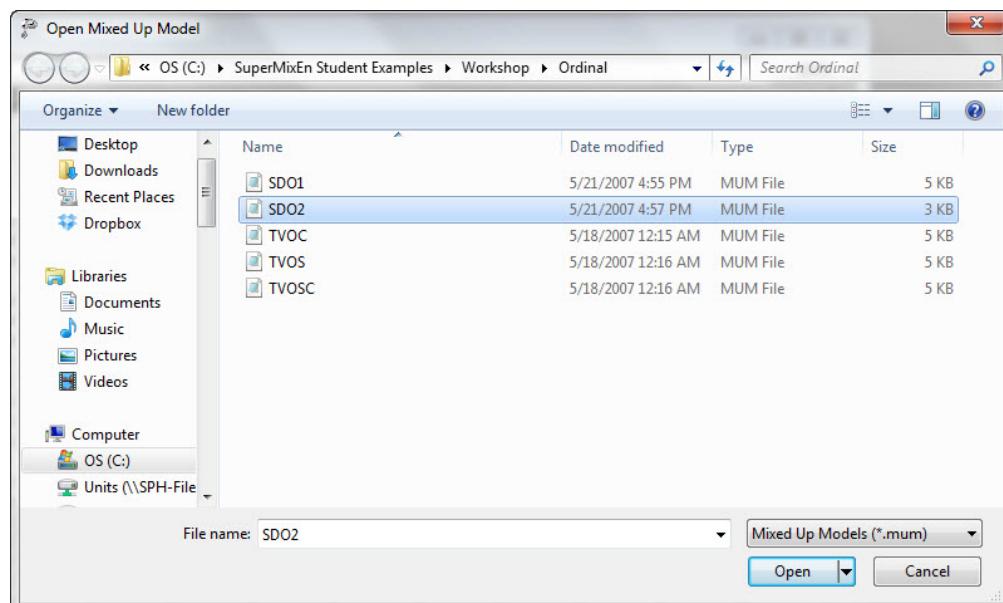
<sup>1</sup> = independent + community vs street

<sup>2</sup> = independent vs community + street

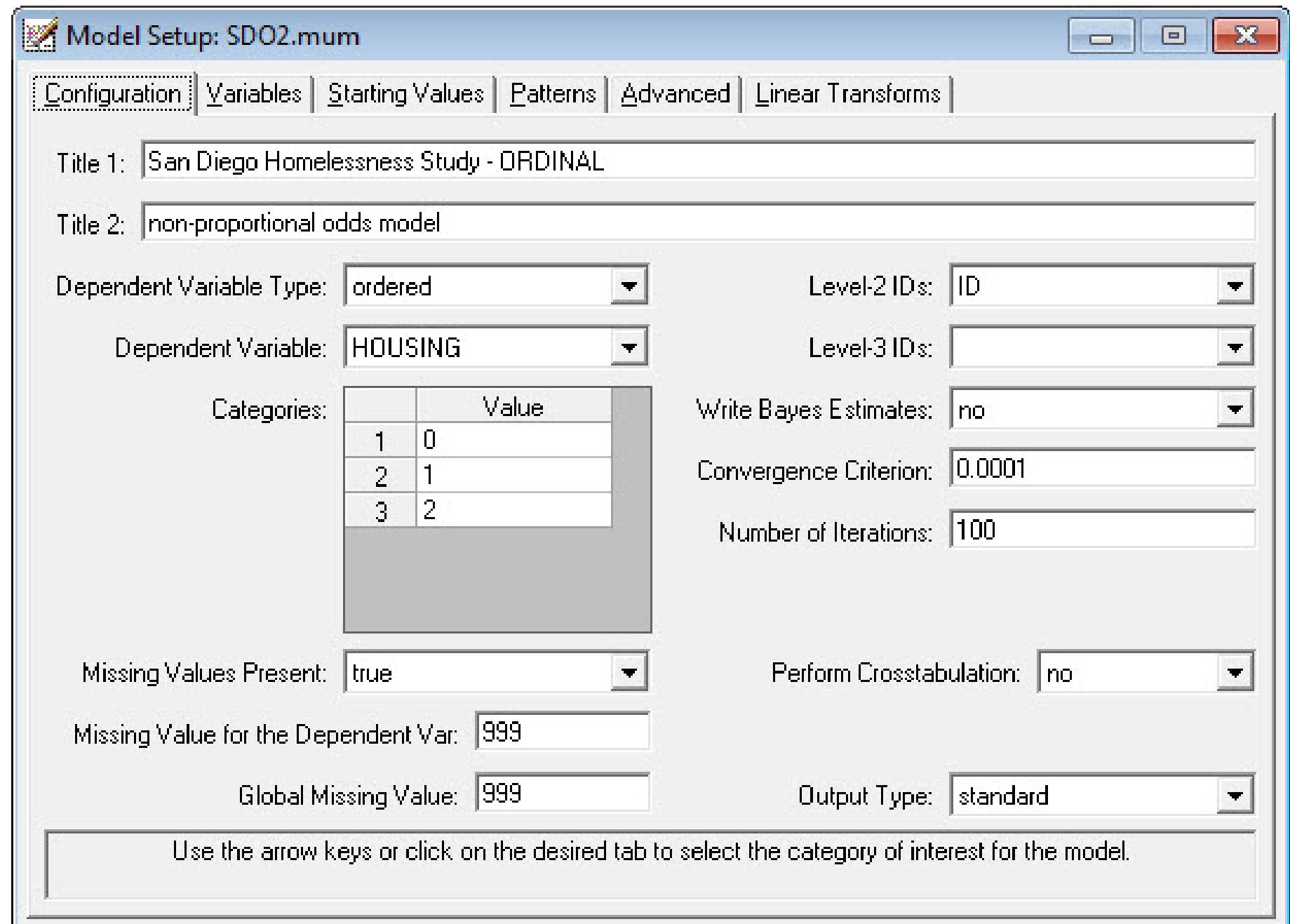
For Non-Proportional Odds model, under “File” click on “Open Existing Model Setup”



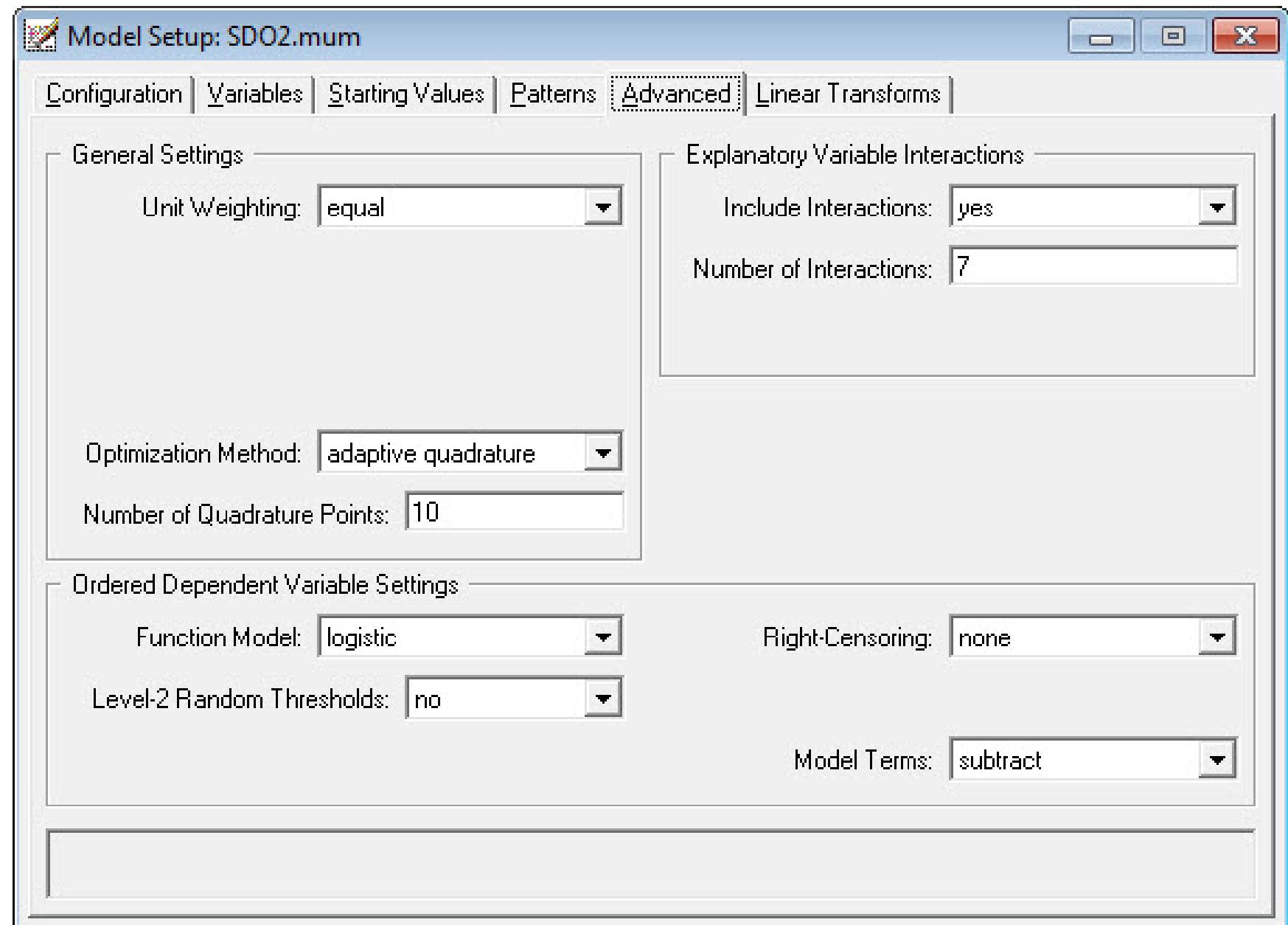
Open C:\SuperMixEn Examples\Workshop\Ordinal\SDO2.mum  
(or C:\SuperMixEn Student Examples\Workshop\Ordinal\SDO2.mum)



Note that “Dependent Variable Type” is “ordered”



Note “Explanatory Variable Interactions” is set to 7



**Model Setup: SDO2.mum**

- [Configuration](#)
- [Variables](#)
- [Starting Values](#)
- [Patterns](#)
- [Advanced](#)
- [Linear Transforms](#)

Linear Transforms	
Time1 Thresh2	<a href="#">Add Transform</a>
	<a href="#">Copy Transform</a>
	<a href="#">Remove Transform</a>

Explanatory Variables:		Level-2 Random Effect (Co)variances:	
	Value		Value
TIME1	1	intercept variance	
TIME2			
TIME3			
SECT8T1			

Thresholds:		Threshold Interactions:	
	Value		Value
1		TIME1	1
2		TIME2	
		TIME3	
		SECT8T1	

[Deletes the currently selected transform: Time1 Thresh2.](#)

**Model Setup: SDO2.mum**

- [Configuration](#)
- [Variables](#)
- [Starting Values](#)
- [Patterns](#)
- [Advanced](#)
- [Linear Transforms](#)

Linear Transforms	
Time1 Thresh2	<a href="#">Add Transform</a>
Time2 Thresh2	<a href="#">Copy Transform</a>
	<a href="#">Remove Transform</a>

Explanatory Variables:		Level-2 Random Effect (Co)variances:	
	Value		Value
TIME1		intercept variance	
TIME2	1		
TIME3			
SECT8T1			

Thresholds:		Threshold Interactions:	
	Value		Value
1		TIME1	
2		TIME2	1
		TIME3	
		SECT8T1	

[Enter Threshold Interactions for the transform Time2 Thresh2.](#)

**Model Setup: SDO2.mum**

- [Configuration](#)
- [Variables](#)
- [Starting Values](#)
- [Patterns](#)
- [Advanced](#)
- [Linear Transforms](#)

Linear Transforms	
Time3 Thresh2	<a href="#">Add Transform</a>
Sect8T1	<a href="#">Copy Transform</a>
Sect8T2	
Sect8T3	<a href="#">Remove Transform</a>

Explanatory Variables:		Level-2 Random Effect (Co)variances:	
	Value		Value
SECT8T1		intercept variance	
SECT8T2			
SECT8T3	1		
SECTION8			

Thresholds:		Threshold Interactions:	
	Value		Value
1		TIME3	
2		SECT8T1	
		SECT8T2	
		SECT8T3	1

[Enter Threshold Interactions for the transform Sect8T3.](#)

**Model Setup: SDO2.mum**

- [Configuration](#)
- [Variables](#)
- [Starting Values](#)
- [Patterns](#)
- [Advanced](#)
- [Linear Transforms](#)

Linear Transforms	
Sect8T1	<a href="#">Add Transform</a>
Sect8T2	<a href="#">Copy Transform</a>
Sect8T3	
Section8	<a href="#">Remove Transform</a>

Explanatory Variables:		Level-2 Random Effect (Co)variances:	
	Value		Value
SECT8T1		intercept variance	
SECT8T2			
SECT8T3			
SECTION8	1		

Thresholds:		Threshold Interactions:	
	Value		Value
1		SECT8T1	
2		SECT8T2	
		SECT8T3	
		SECTION8	1

[Enter Threshold Interactions for the transform Section8.](#)

# Mixed Multinomial Logistic Regression Model

$Y_{ij}$  = nominal response of level-2 unit  $i$  and level-1 unit  $j$

Which member of The Polkaholics is your favorite?  
(asked before, during, and after a show)



# Mixed-effects Multinomial Logistic Regression Model

$$\log \frac{p_{ijc}}{p_{ij1}} = \mathbf{u}'_{ij} \boldsymbol{\gamma}_c + \mathbf{z}'_{ij} \boldsymbol{v}_{ic} \quad c = 2, 3, \dots, C$$

- $C - 1$  contrasts to reference cell ( $c = 1$ )
- regression effects  $\boldsymbol{\gamma}_c$  vary across contrasts
- random-effects  $\boldsymbol{v}_{ic}$  vary across contrasts
  - independent
  - correlated

For example, with  $C = 3$

	contrast	ordinal	nominal
$c1$	2 & 3 vs 1	2 vs 1	
$c2$	3 vs 1 & 2	3 vs 1	

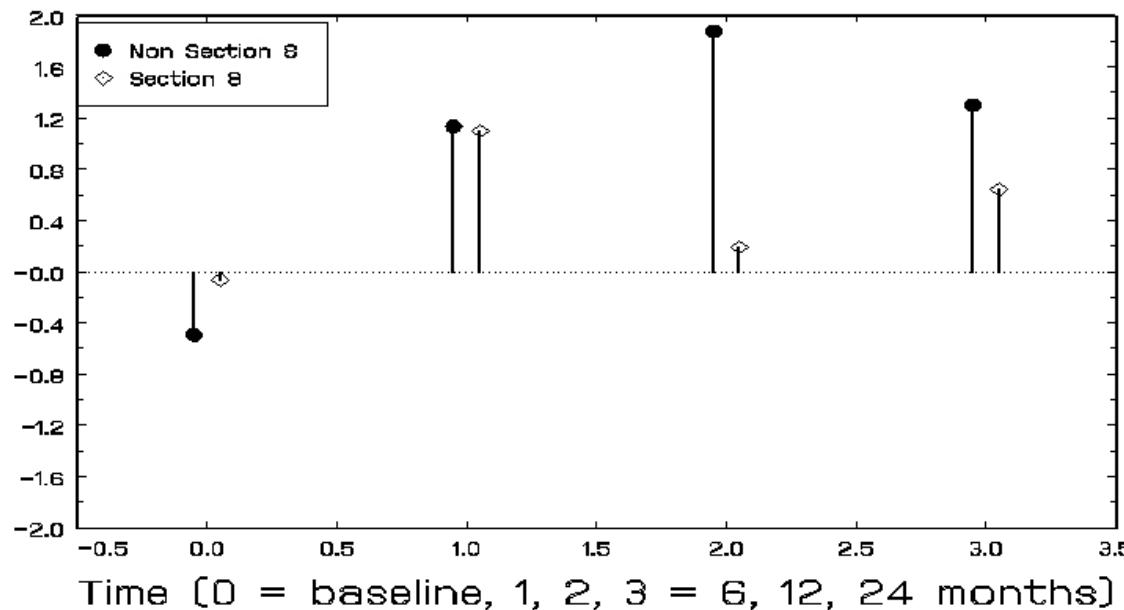
## Model in terms of the category probabilities

$$p_{ijc} = \Pr(Y_{ij} = c \mid \boldsymbol{v}_{ic}) = \frac{\exp(z_{ijc})}{1 + \sum_{h=2}^C \exp(z_{ijh})} \quad \text{for } c = 2, 3, \dots, C$$

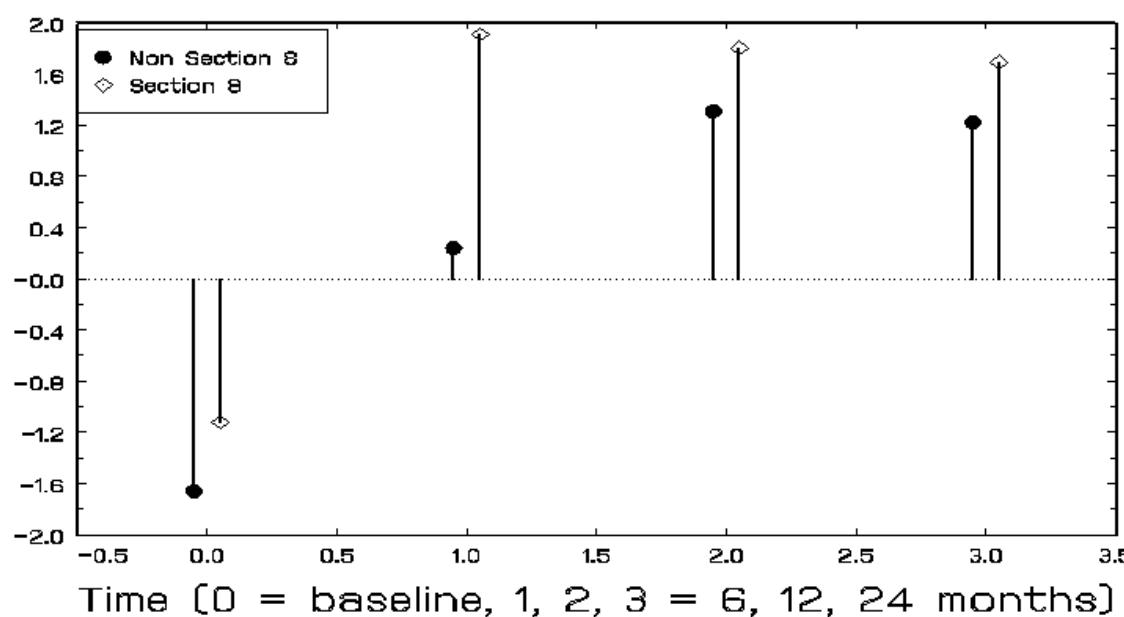
$$p_{ij1} = \Pr(Y_{ij} = 1 \mid \boldsymbol{v}_{ic}) = \frac{1}{1 + \sum_{h=2}^C \exp(z_{ijh})}$$

where the multinomial logit  $z_{ijc} = \boldsymbol{u}'_{ij} \boldsymbol{\gamma}_c + \boldsymbol{z}'_{ij} \boldsymbol{v}_{ic}$

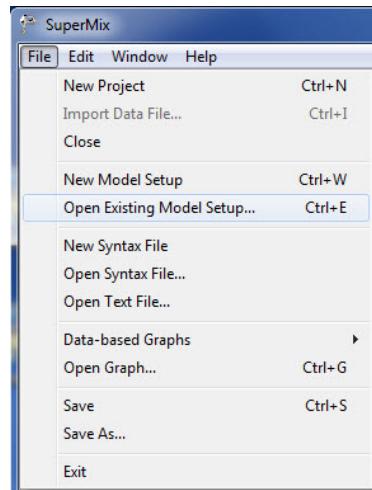
### Empirical Logits - Community vs Street



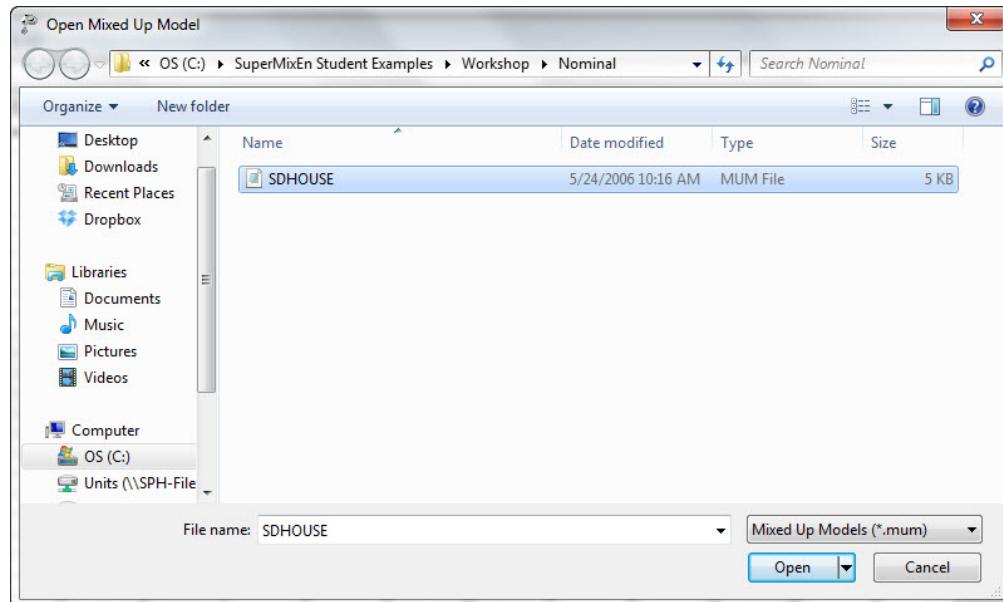
### Empirical Logits - Independent vs Street



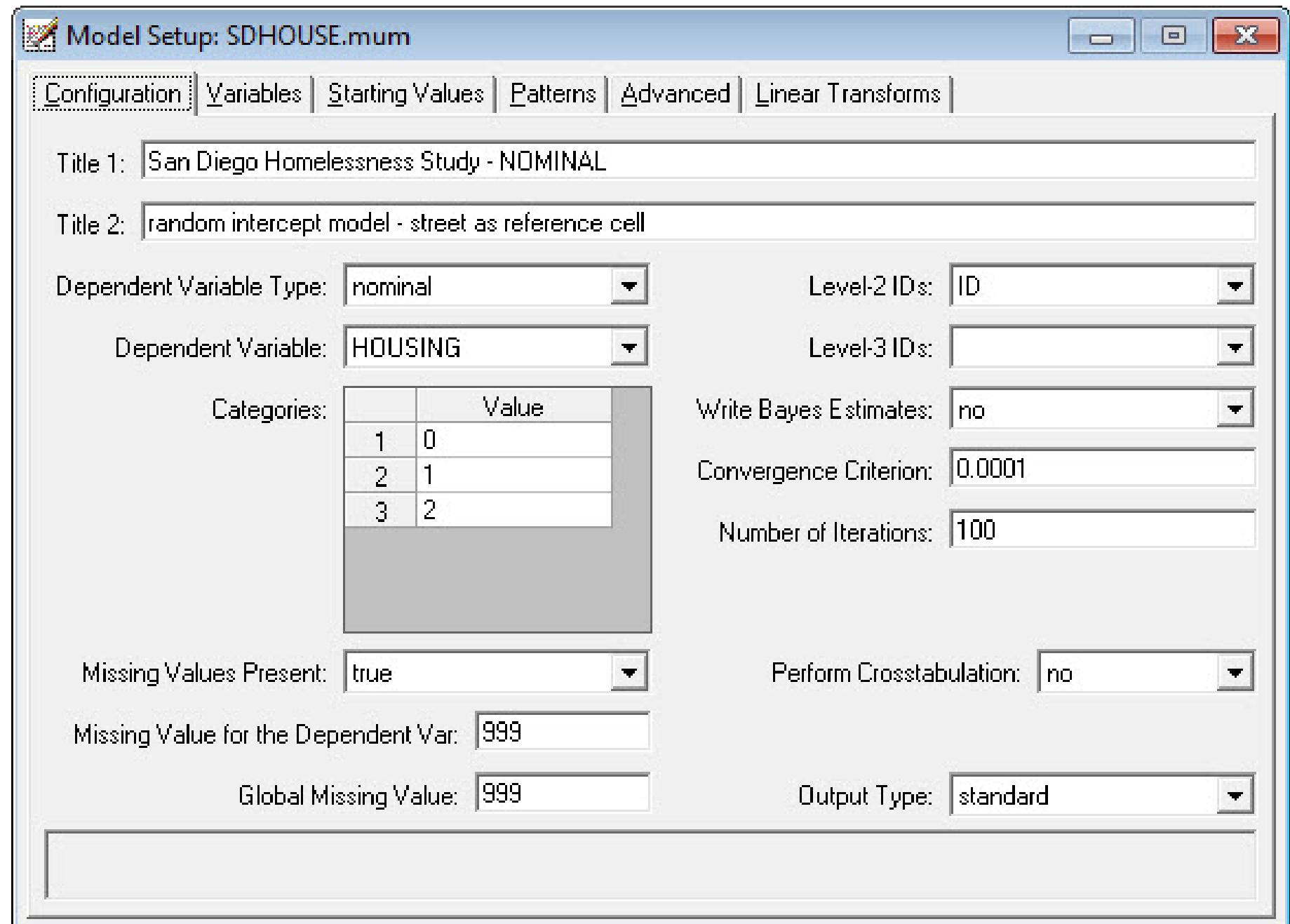
Under “File” click on “Open Existing Model Setup”



Open C:\SuperMixEn Examples\Workshop\Nominal\sdhouse.mum  
(or C:\SuperMixEn Student Examples\Workshop\Nominal\sdhouse.mum)



Note that “Dependent Variable Type” is “nominal”





## Model Setup: SDHOUSE.mum



Configuration

Variables

Starting Values

Patterns

Advanced

Linear Transforms

Available	E	2
ID	<input type="checkbox"/>	<input checked="" type="checkbox"/>
HOUSING	<input type="checkbox"/>	<input checked="" type="checkbox"/>
SECTION8	<input checked="" type="checkbox"/>	<input type="checkbox"/>
TIME1	<input checked="" type="checkbox"/>	<input type="checkbox"/>
TIME2	<input checked="" type="checkbox"/>	<input type="checkbox"/>
TIME3	<input checked="" type="checkbox"/>	<input type="checkbox"/>
SECT8T1	<input checked="" type="checkbox"/>	<input type="checkbox"/>
SECT8T2	<input checked="" type="checkbox"/>	<input type="checkbox"/>
SECT8T3	<input checked="" type="checkbox"/>	<input type="checkbox"/>
NOSECT8	<input type="checkbox"/>	<input type="checkbox"/>
TIME	<input type="checkbox"/>	<input type="checkbox"/>
SEC8TIME	<input type="checkbox"/>	<input type="checkbox"/>

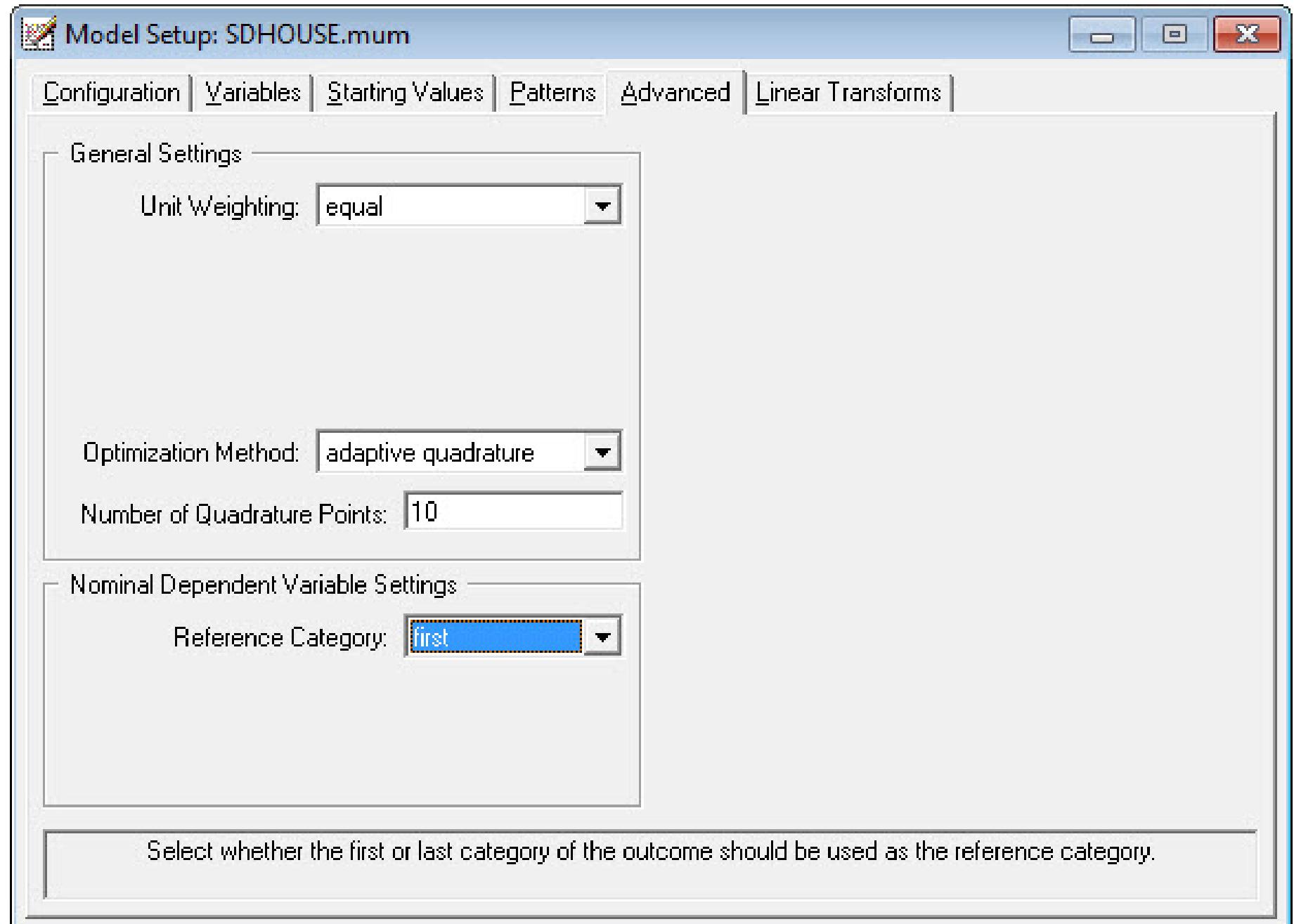
Explanatory Variables
TIME1
TIME2
TIME3
SECTION8
SECT8T1
SECT8T2
SECT8T3

L-2 Random Effects

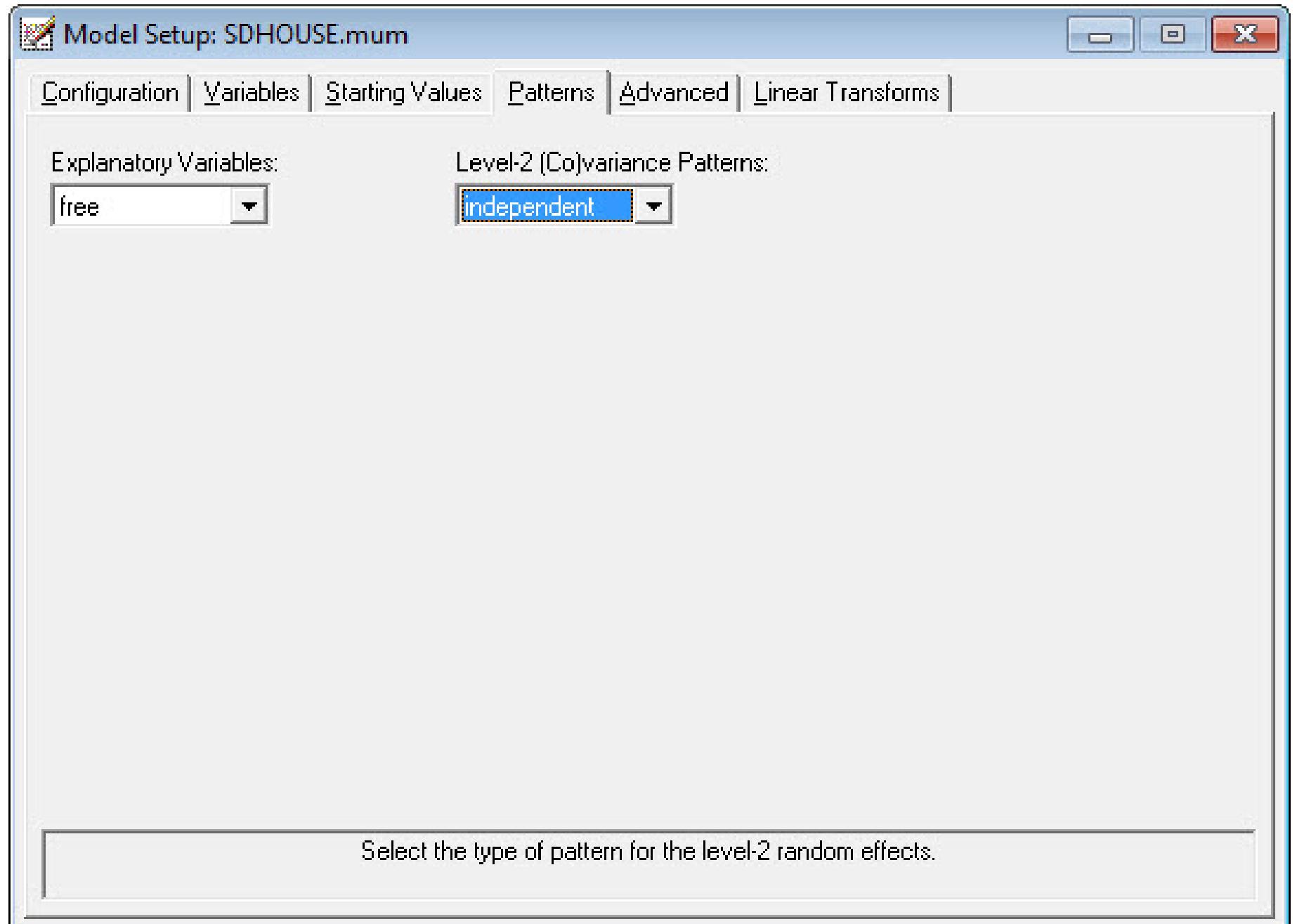
 Include Intercept Include Intercept

Select the columns of the spreadsheet to be used as explanatory variables and random effects.

Can select first or last category as the reference cell



Try independent random effects first



## SDHOUSE.out

```
| Optimization Method: Numerical Quadrature |
```

```
Number of quadrature points = 10
Number of free parameters = 18
Number of iterations used = 5

-2lnL (deviance statistic) = 2218.71755
Akaike Information Criterion 2254.71755
Schwarz Criterion 2347.62674
```

## Estimated regression weights

Parameter	Estimate	Standard Error	z Value	P Value
-----	-----	-----	-----	-----
Response Code 1 vs Code 0				
intercept	-0.4511	0.1847	-2.4425	0.0146
TIME1	1.9421	0.3002	6.4685	0.0000
TIME2	2.8207	0.3838	7.3502	0.0000
TIME3	2.2604	0.3548	6.3709	0.0000
SECTION8	0.5207	0.2618	1.9892	0.0467
SECT8T1	-0.1350	0.4617	-0.2925	0.7699
SECT8T2	-1.9180	0.5167	-3.7122	0.0002
SECT8T3	-0.9525	0.4808	-1.9811	0.0476
Response Code 2 vs Code 0				
intercept	-2.6746	0.3791	-7.0560	0.0000
TIME1	2.6847	0.4362	6.1550	0.0000
TIME2	4.0907	0.4976	8.2208	0.0000
TIME3	4.1009	0.4701	8.7231	0.0000
SECTION8	0.7816	0.4817	1.6225	0.1047
SECT8T1	2.0009	0.6193	3.2311	0.0012
SECT8T2	0.5460	0.6509	0.8389	0.4015
SECT8T3	0.3035	0.6198	0.4896	0.6244

Save As...

Close

## SDHOUSE.out

## Estimated level 2 variances and covariances

Parameter	Estimate	Standard Error	z Value	P Value
intercept1/intercept1	0.7617	0.3280	2.3219	0.0202
intercept2/intercept2	5.4514	1.0237	5.3255	0.0000

## Level 2 covariance matrix

	intercept1	intercept2
intercept1	0.761654	
intercept2	0.000000	5.451441

## Level 2 correlation matrix

	intercept1	intercept2
intercept1	1.000000	
intercept2	0.000000	1.000000

## Calculation of intracluster correlation (01)

```
-----
residualvariance = pi*pi / 3 (assumed)
cluster variance = 0.7617

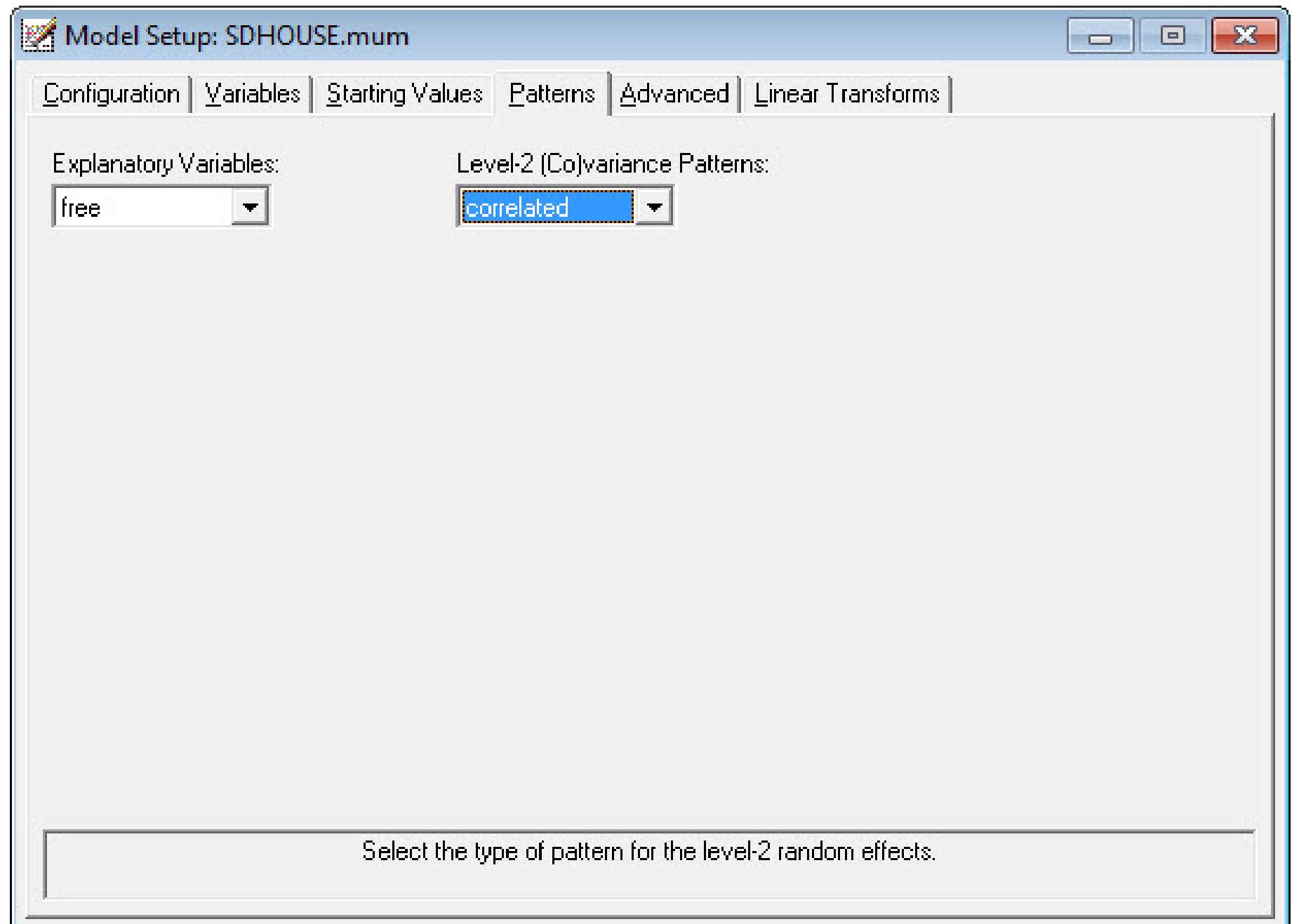
intracluster correlation = 0.7617 / ( 0.7617 + (pi*pi/3)) = 0.188
```

## Calculation of intracluster correlation (02)

```
-----
residualvariance = pi*pi / 3 (assumed)
cluster variance = 5.4514

intracluster correlation = 5.4514 / ( 5.4514 + (pi*pi/3)) = 0.624
```

Now allow the random effects to be correlated



SDHOUSE.out

```

o-----o
| Optimization Method: Adaptive Quadrature |
o-----o

Number of quadrature points =          10
Number of free parameters =           19
Number of iterations used =            7

-2lnL (deviance statistic) =      2180.93648
Akaike Information Criterion       2218.93648
Schwarz Criterion                 2317.00729

Estimated regression weights

      Standard
Parameter   Estimate    Error   z Value   P Value
-----
Response Code 1 vs Code 0
-----
intercept     -0.6222    0.2330  -2.6701  0.0076
TIME1          2.3744    0.3477  6.8292  0.0000
TIME2          3.3452    0.4416  7.5759  0.0000
TIME3          2.5894    0.4022  6.4378  0.0000
SECTION8       0.6516    0.3281  1.9858  0.0471
SECT8T1        -0.3311   0.5228  -0.6334  0.5265
SECT8T2        -2.4746   0.5897  -4.1967  0.0000
SECT8T3        -1.1596   0.5438  -2.1322  0.0330

Response Code 2 vs Code 0
-----
intercept     -2.5987    0.3837  -6.7729  0.0000
TIME1          2.8878    0.4541  6.3591  0.0000
TIME2          4.3939    0.5301  8.2887  0.0000
TIME3          4.3104    0.4958  8.6944  0.0000
SECTION8       0.8314    0.4938  1.6836  0.0923
SECT8T1        1.9688    0.6436  3.0592  0.0022
SECT8T2        0.2872    0.6769  0.4243  0.6713
SECT8T3        0.2017    0.6442  0.3130  0.7543

```

Save As... Close

LR test comparing models:  $\chi^2_1 = 2218.72 - 2180.94 = 37.78$

## SDHOUSE.out

## Estimated level 2 variances and covariances

Parameter	Estimate	Standard Error	z Value	P Value
intercept1/intercept1	2.7025	0.6968	3.8786	0.0001
intercept2/intercept1	2.8846	0.7649	3.7710	0.0002
intercept2/intercept2	5.8229	1.1358	5.1269	0.0000

## Level 2 covariance matrix

	intercept1	intercept2
intercept1	2.702462	
intercept2	2.884597	5.822889

## Level 2 correlation matrix

	intercept1	intercept2
intercept1	1.000000	
intercept2	0.727170	1.000000

## Calculation of intracluster correlation (01)

-----  
residualvariance = pi\*pi / 3 (assumed)  
cluster variance = 2.7025  
  
intracluster correlation = 2.7025 / ( 2.7025 + (pi\*pi/3)) = 0.451

## Calculation of intracluster correlation (02)

-----  
residualvariance = pi\*pi / 3 (assumed)  
cluster variance = 5.8229  
  
intracluster correlation = 5.8229 / ( 5.8229 + (pi\*pi/3)) = 0.639

Save As...

Close

# Summary

Models for longitudinal ordinal and nominal data as developed as models for continuous and dichotomous data

- Proportional odds models
- Non and partial proportional odds models
- Nominal models (with reference-cell contrasts)
- Grouped-time survival analysis models

⇒ SuperMix can do it all, including 3-level models, also for counts

- full likelihood solution using adaptive quadrature