Causal Modeling With GSS Data Using Multiple Approaches

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Setup

- . clear all
- . cd "/Users/agrogan/Desktop/newstuff/causal-modeling" /Users/agrogan/Desktop/newstuff/causal-modeling

Get Data

```
. use "/Users/agrogan/Box Sync/DATA WAREHOUSE/General Social Survey Panel Data/GSS_panel2 > 010w123_R6 - stata.dta", clear
```

ID Variable

. generate ID = id_1

Keep Only Relevant Variables

. keep ID satjob_? educ_? race_? incom16_?

Describe Data

. describe

Contains data from /Users/agrogan/Box Sync/DATA WAREHOUSE/General Social Survey Panel Dat > $a/GSS_panel2010w123_R6 - stata.dta$

obs: 2,044 vars: 13 size: 32,704

12 MAR 2018 16:24

variable name	storage type	display format	value label	variable label
educ_1	byte	%8.0g	EDUC_1	educ_1: HIGHEST YEAR OF SCHOOL COMPLETED
educ_2	byte	%8.0g	EDUC_2	educ_2: HIGHEST YEAR OF SCHOOL COMPLETED
educ_3	byte	%8.0g	EDUC_3	educ_3: HIGHEST YEAR OF SCHOOL COMPLETED
incom16_1	byte	%8.0g	INCOM16	incom16_1: RS FAMILY INCOME WHEN 16 YRS OLD
incom16_2	byte	%8.0g	V1318_A	incom16_2: RS FAMILY INCOME WHEN 16 YRS OLD
incom16_3	byte	%8.0g	V1319_A	incom16_3: RS FAMILY INCOME WHEN 16 YRS OLD
race_1	byte	%8.0g	RACE_1	race_1: RACE OF RESPONDENT
race_2	byte	%8.0g	RACE_2	race_2: RACE OF RESPONDENT
race_3	byte	%8.0g	RACE_3	race_3: RACE OF RESPONDENT
satjob_1	byte	%8.0g	SATJOB_1	satjob_1: JOB OR HOUSEWORK

satjob_2	byte	%8.0g	SATJOB_2	satjob_2:	JOB OR HOUSE	JORK
satjob_3	byte	%8.0g	SATJOB_3	satjob_3:	JOB OR HOUSE	NORK
ID	float	%9.0g				

Sorted by:

Note: Dataset has changed since last saved.

Codebook for Selected Variable(s)

. codebook satjob_3

satjob_3: JOB OR HOUSEWORK satjob_3 type: numeric (byte) label: SATJOB_3 range: [1,4] units: 1 missing .: 0/2,044 unique values: 4 unique mv codes: 3 missing .*: 1,086/2,044 tabulation: Freq. Numeric Label VERY SATISFIED 483 1 2 MOD. SATISFIED 69 3 A LITTLE DISSAT 39 4 VERY DISSATISFIED 4 .d DK 1,073 .i IAP .n NA

Analyses Relying On Wide Data

Correlation

Regression with 1 Independent Variable

. regress satjob_3 educ_3 Source df MS Number of obs 957 F(1, 955) 5.76 Model 3.53828635 3.53828635 Prob > F 0.0166 R-squared Residual 586.493062 955 .61412886 0.0060 Adj R-squared 0.0050 Total 590.031348 956 .617187602 Root MSE .78366 satjob_3 Coef. Std. Err. t P>|t| [95% Conf. Interval] educ_3 -.0216864 .0090349 -2.40 0.017 -.0394169 -.003956 _cons 1.954439 .1297867 15.06 0.000 1.699739 2.209139

Regression With Multiple Independent Variables

		_	_	_		_		_
regress	satiob	3	educ	3	i.race	3	incom16	3

Source	SS	df	MS	Number of obs	=	951
				F(4, 946)	=	2.36
Model	5.81703392	4	1.45425848	Prob > F	=	0.0517
Residual	582.580442	946	.615835563	R-squared	=	0.0099
				Adj R-squared	=	0.0057
Total	588.397476	950	.619365765	Root MSE	=	.78475
satjob_3	Coef.	Std. Err.	t]	P> t [95% Co	nf.	Interval]
educ_3	0215151	.0092674	-2.32	0.020039702	1	0033281
race 3						
black	.1267666	.0708898	1.79	0.074012352	8	.2658861
other	.0677238	.0985112	0.69	0.492125601	9	.2610495
incom16_3	.0115275	.0280601	0.41	0.681043539	8	.0665947
_cons	1.89556	.144649	13.10	0.000 1.6116	9	2.17943

Propensity Score

Data Wrangling

- . generate twelve_years_3 = educ_3 >= 12 // 12 or more years of education
- . generate twelve_years_2 = educ_2 >= 12 // 12 or more years of education
- . generate twelve_years_1 = educ_1 >= 12 // 12 or more years of education

Propensity Score Analysis

satjob_3	Coef.	AI Robust Std. Err.	z	P> z	[95% Conf.	Interval]
ATE twelve_years_3 (1 vs 0)	0410168	.1083808	-0.38	0.705	2534393	.1714057

Check Balance

Analyses Relying On Long Data

Reshape The Data

. reshape long satjob_ educ_ twelve_years_ incom16_ race_, i(ID) j(wave) (note: $j = 1 \ 2 \ 3$)

Data	wide	->	long
Number of obs.	2044	->	6132

```
Number of variables
                                    16
j variable (3 values)
                                         ->
                                              wave
xij variables:
                                         ->
            satjob_1 satjob_2 satjob_3
                                              satjob_
                  educ_1 educ_2 educ_3
                                              educ_
twelve_years_1 twelve_years_2 twelve_years_3-> twelve_years_
         incom16_1 incom16_2 incom16_3
                                        ->
                                              incom16_
                  race_1 race_2 race_3
                                              race_
```

Clean Up Variable Names

```
. rename satjob_ satjob
```

- . rename educ_ educ
- . rename incom16_ incom16
- . rename race_ race
- . rename twelve_years_ twelve_years

Multilevel Model

```
. mixed satjob wave educ incom16 i.race || ID:
Performing EM optimization:
Performing gradient-based optimization:
Iteration 0: log likelihood = -4161.775
               log likelihood = -4161.7476
log likelihood = -4161.7476
Iteration 1:
Iteration 2:
Computing standard errors:
Mixed-effects ML regression
                                                   Number of obs
Group variable: ID
                                                   Number of groups =
                                                   Obs per group:
                                                                  avg =
                                                                  max =
                                                   Wald chi2(5)
Log likelihood = -4161.7476
                                                   Prob > chi2
```

satjob	Coef.	Std. Err.	Z	P> z	[95% Conf.	Interval]
wave	018625	.014015	-1.33	0.184	0460938	.0088439
educ	018976	.0054133	-3.51	0.000	0295859	008366
incom16	0350535	.0154559	-2.27	0.023	0653465	0047606
race						
black	.1695589	.0451171	3.76	0.000	.0811311	.2579868
other	.035975	.0543135	0.66	0.508	0704776	.1424276
_cons	2.049073	.0843019	24.31	0.000	1.883845	2.214302

Random-effe	cts Parameters	Estimate	Std. Err.	[95% Conf.	Interval]
ID: Identity	var(_cons)	.2305185	.0161162	. 2009999	.2643722
	var(Residual)	.4174209	.0131143	.3924927	.4439323

LR test vs. linear model: chibar2(01) = 322.95

Prob >= chibar2 = 0.0000

3,595

1,661

1

2.2

42.38

0.0000

Fixed effects regression

. xtreg satjob wave educ incom16 i.race, i(ID) fe							
Fixed-effects	(within) regr	ression		Number	of obs =	3,595	
Group variable	e: ID			Number	of groups =	1,661	
R-sq:				Obs per	group:		
within =	= 0.0052				min =	1	
between =	= 0.0148				avg =	2.2	
overall =	= 0.0122				max =	3	
				F(5,192	9) =	2.03	
corr(u_i, Xb)	= -0.0714			Prob > 1	F =	0.0711	
satjob	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]	
wave	0237842	.0152551	-1.56	0.119	0537023	.006134	
educ	0087664	.0158008	-0.55	0.579	0397548	.022222	
incom16	047186	.0228265	-2.07	0.039	0919531	0024189	
race							
black	.3226033	.2025604	1.59	0.111	0746572	.7198637	
other	.0383663	.104807	0.37	0.714	1671806	.2439132	
_cons	1.928458	. 227991	8.46	0.000	1.481323	2.375593	
sigma_u	.6861769						
sigma_e	.64822634						
rho	.52841711	(fraction	of varia	nce due to	o u_i)		

F test that all $u_i=0$: F(1660, 1929) = 2.18

Prob > F = 0.0000

"Hybrid" Model

Generate Within And Between Variables

```
. bysort ID: egen educ_mean = mean(educ)
(6 missing values generated)
```

. generate educ_deviation = educ - educ_mean
(1,240 missing values generated)

. mixed satjob wave educ_mean educ_deviation incom16 i.race || ID:

Performing EM optimization:

Performing gradient-based optimization:

Iteration 0: log likelihood = -4161.3224
Iteration 1: log likelihood = -4161.2951
Iteration 2: log likelihood = -4161.2951

Computing standard errors:

Mixed-effects ML regression	Number of obs	=	3,595
Group variable: ID	Number of groups	; =	1,661
	Obs per group:		
	mi	n =	1
	ar	7g =	2.2
	ma	ax =	3
	Wald chi2(6)	=	43.30
Log likelihood = -4161.2951	Prob > chi2	=	0.0000

satjob	Coef.	Std. Err.	z	P> z	[95% Conf.	Interval]
wave educ_mean educ_deviation incom16	0197009 0208983 0054971 0343579	.0140588 .0057775 .0151667 .0154712	-1.40 -3.62 -0.36 -2.22	0.161 0.000 0.717 0.026	0472556 0322221 0352233 0646809	.0078537 0095745 .0242292 0040349
race						

black	.1684699	.0451261		0.000	.0800245	.2569154
other	.0342568	.0543368		0.528	0722414	.140755
_cons	2.075849	.088866	23.36	0.000	1.901675	2.250023

Random-effects Parameters		Estimate	Std. Err.	[95% Conf.	Interval]
ID: Identity					
	var(_cons)	.2304651	.0161097	.2009581	.2643046
	var(Residual)	.4173132	.0131099	.3923934	.4438157

LR test vs. linear model: chibar2(01) = 323.08 Prob >= chibar2 = 0.0000