Note: Printing Stata output is annoying in that you can only print so many lines at once, so you have to print long runs in multiple parts then stitch the pdfs together. The issue with this is that each part can include the tail end of the output of previous parts. This means that the left-hand line numbers are incorrect and that certain outputs twice appear in two subsequent parts (once at the end of a page, then again at the top of the next page). To reduce confusion, we've blacked out any repeated outputs to reproduce the full, final Stata run from start to finish.

  ___ ____ ____ ____ ____(R)
 /__   /   ____/   /   ____/
___/   /   /___/   /   /___/
  Statistics/Data analysis

  ___ ____ ____ ____ ____ ®
 /__   /   ____/   /   ____/   StataNow 19.5
___/   /   /___/   /   /___/   SE—Standard Edition

 Statistics and Data Science   Copyright 1985-2025 StataCorp LLC
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  Notes:
        1. Unicode is supported; see help unicode_advice.
        2. Maximum number of variables is set to 32,767; see help set_maxvar.
        3. New update available; type -update all-

```
 1 . doedit "C:\Users\14258\Dropbox\Projects\Auth COVID Replication\Analysis\Reproduction Files\main_analysis.do"

 2 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000000.tmp"

 3 . /*******************************************************************
   >   main_analysis.do
   >   Authors: Adam Panish*, Trent Ollerenshaw, and Joseph Vitriol
   >   Note: Put this do-file in the same folder as pooled_data.xlsx
   >
   > *******************************************************************/
 4 .
 5 . ***** Session Setup *****
 6 . clear all

 7 .
 8 . * Import Dataset (make sure it's in the same folder as the do file!)
 9 . import excel "pooled_data.xlsx", sheet("Sheet1") firstrow clear
   (190 vars, 10,223 obs)

10 .
11 . set seed 1996 // for reproducibility

12 .
13 . * Standardize Some Variables
14 . local vars dv_concern dv_behavior dv_restrictions auth_scale latentnsc1 latentnsc2 polengage_scale polint_scale polkno

15 . foreach var of local vars {
     2.     egen mean_`var' = mean(`var')
     3.     egen sd_`var' = sd(`var')
     4.     generate z_`var' = (`var' - mean_`var') / sd_`var'
     5. }
   (3,740 missing values generated)
   (5,219 missing values generated)
   (3,352 missing values generated)
   (1,385 missing values generated)
   (9,177 missing values generated)
   (9,177 missing values generated)
   (282 missing values generated)
   (1,229 missing values generated)
   (935 missing values generated)
```

```
16 .
17 . ***** Table 1 Descriptive Statistics *****
18 .
19 . * Sample Sizes (useable observations only)
20 . tab sample if (auth_scale != . | latentnsc2 != .) & polengage_scale != . & (dv_concern != . | dv_restrictions != . | c
```

sample	Freq.	Percent	Cum.
ANESGSS	**775**	**8.92**	**8.92**
Bovitz	**496**	**5.71**	**14.63**
Lucid2020	**1,007**	**11.59**	**26.22**
Lucid2021	**2,119**	**24.39**	**50.62**
MTurk	**546**	**6.29**	**56.90**
VSG	**3,744**	**43.10**	**100.00**
Total	**8,687**	**100.00**	

```
21 .
22 . * Alphas
23 . alpha auth1-auth4 if sample == "ANESGSS"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0717019
   Number of items in the scale:            4
   Scale reliability coefficient:      0.6629

24 . alpha auth1-auth5 if sample == "Lucid2020"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0842526
   Number of items in the scale:            5
   Scale reliability coefficient:      0.7378

25 . alpha auth1-auth8 if sample == "Lucid2021"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0604484
   Number of items in the scale:            8
   Scale reliability coefficient:      0.7423

26 . alpha auth1-auth4 if sample == "VSG"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0817513
   Number of items in the scale:            4
   Scale reliability coefficient:      0.6883

27 . alpha auth1-auth4 if sample == "MTurk"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0454346
   Number of items in the scale:            4
   Scale reliability coefficient:      0.7584
```

```
28 . alpha auth1-auth4 if sample == "Bovitz"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0254874
   Number of items in the scale:            4
   Scale reliability coefficient:      0.6240

29 .
30 . alpha polint1-polint2 if sample == "Lucid2020"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0734526
   Number of items in the scale:            2
   Scale reliability coefficient:      0.8320

31 . alpha polint1-polint2 if sample == "Lucid2021"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0784034
   Number of items in the scale:            2
   Scale reliability coefficient:      0.8296

32 . alpha polint1-polint6 if sample == "VSG"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0499288
   Number of items in the scale:            6
   Scale reliability coefficient:      0.9554

33 . alpha polint1-polint3 if sample == "MTurk"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0350215
   Number of items in the scale:            3
   Scale reliability coefficient:      0.7565

34 . alpha polint1-polint3 if sample == "Bovitz"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0431443
   Number of items in the scale:            3
   Scale reliability coefficient:      0.7523

35 .
36 . alpha polknow4-polknow8 if sample == "ANESGSS" // note: the first three knowledge items are pre-election, so missing

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0483769
   Number of items in the scale:            5
   Scale reliability coefficient:      0.6731
```

```
37 . alpha polknow1-polknow6 if sample == "Lucid2020"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0610431
   Number of items in the scale:            6
   Scale reliability coefficient:      0.7360

38 . alpha polknow1-polknow5 if sample == "Lucid2021"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0677815
   Number of items in the scale:            5
   Scale reliability coefficient:      0.7081

39 . alpha polknow1-polknow7 if sample == "VSG"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0594696
   Number of items in the scale:            7
   Scale reliability coefficient:      0.7892

40 . alpha polknow1-polknow8 if sample == "MTurk"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0413857
   Number of items in the scale:            8
   Scale reliability coefficient:      0.7557

41 . alpha polknow1-polknow8 if sample == "Bovitz"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0539215
   Number of items in the scale:            8
   Scale reliability coefficient:      0.7570

42 .
43 . alpha dv_masking dv_distancing if sample == "Lucid2020"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0395013
   Number of items in the scale:            2
   Scale reliability coefficient:      0.8102

44 . alpha dv_masking dv_distancing dv_vaxstatus if sample == "Lucid2021"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0384375
   Number of items in the scale:            3
   Scale reliability coefficient:      0.5661
```

```
45 . alpha dv_masking dv_distancing dv_travel-dv_sanitize if sample == "MTurk"

   Test scale = mean(unstandardized items)

   Average interitem covariance:      .032979
   Number of items in the scale:            8
   Scale reliability coefficient:      0.7522

46 . alpha dv_masking dv_distancing dv_travel-dv_sanitize if sample == "Bovitz"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0351507
   Number of items in the scale:            8
   Scale reliability coefficient:      0.7427

47 .
48 . alpha dv_businesses dv_tracking dv_lockdowns dv_maskmandates if sample == "Lucid2020"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0534972
   Number of items in the scale:            4
   Scale reliability coefficient:      0.8124

49 . alpha dv_businesses dv_tracking dv_lockdowns dv_maskmandates dv_vaxmandate if sample == "Lucid2021"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0738588
   Number of items in the scale:            5
   Scale reliability coefficient:      0.8922

50 . alpha dv_businesses dv_staterestrict-dv_testingreq if sample == "VSG"

   Test scale = mean(unstandardized items)

   Average interitem covariance:     .0965356
   Number of items in the scale:            8
   Scale reliability coefficient:      0.9430

51 .
52 . * 5th to 95th Percentiles of Authoritarianism and Latent NSC
53 . sum auth_scale, det // 0 to 1

                          auth_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%           0              0
 5%           0              0
10%           0              0       Obs               8,838
25%         .25              0       Sum of wgt.       8,838

50%          .5                      Mean           .4921391
                      Largest       Std. dev.       .320342
75%         .75              1
90%           1              1       Variance        .102619
95%           1              1       Skewness      -.0059802
99%           1              1       Kurtosis       1.951514
```

```
54 . sum latentnsc2, det // 0.055 to 0.782

                          latentnsc2
-------------------------------------------------------------
      Percentiles      Smallest
 1%     .0155201              0
 5%     .0553126       .0043632
10%     .1453914       .0071583       Obs               1,046
25%     .3387757       .0073306       Sum of wgt.       1,046

50%     .4893718                      Mean           .4555281
                        Largest       Std. dev.      .2029025
75%     .5534478       .9438711
90%     .7083806       .9544292       Variance       .0411694
95%     .7823676       .9549724       Skewness      -.2249455
99%     .9129034              1       Kurtosis       2.952598

55 .
56 . ***** SM1 - Key Variable Histograms *****
57 . histogram auth_scale, percent bin(4) by(sample, note("") col(3)) title("")

58 . histogram latentnsc2, percent bin(6) by(sample, note("") col(3)) title("")

59 . histogram polengage_scale, percent bin(6) by(sample, note("") col(3)) title("")

60 . histogram dv_concern, percent bin(3) by(sample, note("") col(3)) title("")

61 . histogram dv_restrictions, percent bin(4) by(sample, note("") col(3)) title("")

62 . histogram dv_behavior, percent bin(3) by(sample, note("") col(3)) title("")

63 .
64 .
65 . ***** Main Text Analyses & SM3 Regression Tables *****
66 .
67 . * Set Up Multiple Imputation
68 . mi set mlong

69 . mi register imputed age education income male race sample_num
   (1198 m=0 obs now marked as incomplete)

70 . mi impute chained (logit) male (mlogit) race sample_num (regress) age education income, add(25)
   note: variable sample_num contains no soft missing (.) values; imputing nothing

   Conditional models:
            education: regress education i.sample_num i.race i.male age income
                 race: mlogit race i.sample_num education i.male age income
                 male: logit male i.sample_num education i.race age income
                  age: regress age i.sample_num education i.race i.male income
               income: regress income i.sample_num education i.race i.male age

   Performing chained iterations ...

   Multivariate imputation                     Imputations =       25
   Chained equations                                 added =       25
   Imputed: m=1 through m=25                       updated =        0

   Initialization: monotone                     Iterations =      250
                                                   burn-in =       10

                 male: logistic regression
                 race: multinomial logistic regression
           sample_num: multinomial logistic regression
                  age: linear regression
            education: linear regression
               income: linear regression
```

Variable	Observations per *m*			
	Complete	Incomplete	Imputed	Total
male	**9681**	**542**	**542**	**10223**
race	**9898**	**325**	**325**	**10223**
sample_num	**10223**	**0**	**0**	**10223**
age	**9666**	**557**	**557**	**10223**
education	**9926**	**297**	**297**	**10223**
income	**9075**	**1148**	**1148**	**10223**

(Complete + Incomplete = Total; Imputed is the minimum across *m*
 of the number of filled-in observations.)

```
71 .
72 . * COVID-19 Concern, Authoritarianism
73 . mi estimate: regress dv_concern c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.r
   > ##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =      5,484
                                              Average RVI       =     0.0037
                                              Largest FMI       =     0.0614
                                              Complete DF       =       5412
DF adjustment:   Small sample                 DF:     min       =   2,839.85
                                                      avg       =   5,250.52
                                                      max       =   5,406.85
Model F test:       Equal FMI                 F(  71, 5409.7)   =       8.30
Within VCE type:          OLS                 Prob > F          =     0.0000
```

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	**.1698144**	**.0488236**	**3.48**	**0.001**	**.0741006**	**.2655282**
polengage_scale	**.2791857**	**.0989894**	**2.82**	**0.005**	**.0851264**	**.4732451**
c.auth_scale#c.polengage_scale	**-.4280247**	**.0594535**	**-7.20**	**0.000**	**-.5445775**	**-.3114719**
age	**.2139976**	**.074115**	**2.89**	**0.004**	**.0687024**	**.3592929**
polengage_scale	**0**	(omitted)				
c.age#c.polengage_scale	**-.2840209**	**.0953901**	**-2.98**	**0.003**	**-.471024**	**-.0970178**
1.male	**.0042296**	**.032134**	**0.13**	**0.895**	**-.0587659**	**.0672252**
polengage_scale	**0**	(omitted)				
male#c.polengage_scale						
1	**-.0799572**	**.0392459**	**-2.04**	**0.042**	**-.1568951**	**-.0030194**
race						
1	**-.1458162**	**.0588068**	**-2.48**	**0.013**	**-.2611012**	**-.0305311**
2	**.0011709**	**.0675823**	**0.02**	**0.986**	**-.1313178**	**.1336597**
3	**-.0224712**	**.0734298**	**-0.31**	**0.760**	**-.1664235**	**.1214811**
polengage_scale	**0**	(omitted)				
race#c.polengage_scale						
1	**.1194261**	**.0746888**	**1.60**	**0.110**	**-.0269943**	**.2658464**
2	**.0907412**	**.0888589**	**1.02**	**0.307**	**-.0834585**	**.2649408**
3	**.0342578**	**.0953161**	**0.36**	**0.719**	**-.1526006**	**.2211163**
education	**-.0683349**	**.0572198**	**-1.19**	**0.232**	**-.1805094**	**.0438395**
polengage_scale	**0**	(omitted)				
c.education#c.polengage_scale	**.1289616**	**.0703315**	**1.83**	**0.067**	**-.0089172**	**.2668403**

income	-.0681065	.0614003	-1.11	0.267	-.1884949	.0522818
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.0020125	.0755249	0.03	0.979	-.1460767	.1501017
sample_num						
2	-.1528253	.0209469	-7.30	0.000	-.1938909	-.1117598
5	-.1811252	.018906	-9.58	0.000	-.2181887	-.1440616
6	-.0749138	.0146682	-5.11	0.000	-.1036705	-.0461572
state_1	.0054104	.159844	0.03	0.973	-.3079492	.3187699
state_2	-.0337456	.1459829	-0.23	0.817	-.319932	.2524409
state_3	.0731501	.1466206	0.50	0.618	-.2142864	.3605867
state_4	-.0834837	.1421906	-0.59	0.557	-.3622356	.1952682
state_5	.0179579	.140309	0.13	0.898	-.2571053	.2930211
state_6	.0082332	.1436126	0.06	0.954	-.2733064	.2897728
state_7	.0279018	.1480323	0.19	0.851	-.2623022	.3181059
state_8	.1703934	.1704803	1.00	0.318	-.1638174	.5046041
state_9	.083056	.1549788	0.54	0.592	-.2207658	.3868778
state_10	.0073475	.1404176	0.05	0.958	-.2679286	.2826236
state_11	-.0004104	.1414739	-0.00	0.998	-.2777573	.2769366
state_12	.0732503	.1578026	0.46	0.643	-.2361071	.3826077
state_13	.0502092	.1441543	0.35	0.728	-.2323924	.3328107
state_14	-.0111042	.1477027	-0.08	0.940	-.300662	.2784536
state_15	.0343601	.1412601	0.24	0.808	-.2425677	.3112879
state_16	-.0080554	.1429682	-0.06	0.955	-.2883318	.272221
state_17	.0295981	.1455978	0.20	0.839	-.2558334	.3150295
state_18	.0620919	.1452015	0.43	0.669	-.2225627	.3467466
state_19	-.0373625	.1477189	-0.25	0.800	-.3269522	.2522272
state_20	.015938	.1431355	0.11	0.911	-.2646662	.2965423
state_21	.0957577	.1429317	0.67	0.503	-.1844471	.3759624
state_22	.080559	.1509398	0.53	0.594	-.2153447	.3764627
state_23	-.0318926	.141428	-0.23	0.822	-.3091496	.2453645
state_24	-.024006	.1440517	-0.17	0.868	-.3064065	.2583944
state_25	-.0246559	.1428907	-0.17	0.863	-.3047803	.2554685
state_26	.0538343	.1479543	0.36	0.716	-.2362167	.3438852
state_27	-.0486302	.1565793	-0.31	0.756	-.3555896	.2583292
state_28	-.0012074	.1418671	-0.01	0.993	-.2793251	.2769104
state_29	-.1785343	.1963053	-0.91	0.363	-.5633722	.2063037
state_30	-.0015901	.1521536	-0.01	0.992	-.2998733	.2966931
state_31	.0119497	.1521974	0.08	0.937	-.2864194	.3103189
state_32	.0261768	.1420213	0.18	0.854	-.2522432	.3045968
state_33	-.0565228	.1498138	-0.38	0.706	-.3502191	.2371735
state_34	.0004768	.1463846	0.00	0.997	-.286497	.2874506
state_35	.0570415	.140732	0.41	0.685	-.2188509	.332934
state_36	.0446526	.1414324	0.32	0.752	-.2326129	.3219181
state_37	.0971401	.1483953	0.65	0.513	-.1937755	.3880558
state_38	-.0498003	.1445718	-0.34	0.731	-.3332203	.2336197
state_39	.0272937	.1408788	0.19	0.846	-.2488867	.3034742
state_40	.1023423	.159602	0.64	0.521	-.2105429	.4152275
state_41	.0082893	.1448058	0.06	0.954	-.2755895	.2921681
state_42	.0591517	.155159	0.38	0.703	-.2450233	.3633267
state_43	.0015716	.1432004	0.01	0.991	-.27916	.2823032
state_44	-.0098265	.1405358	-0.07	0.944	-.2853345	.2656814
state_45	-.0265837	.1465046	-0.18	0.856	-.3137928	.2606254
state_46	.006442	.1423456	0.05	0.964	-.2726138	.2854978
state_47	.1069835	.163897	0.65	0.514	-.2143214	.4282884
state_48	-.0487794	.142102	-0.34	0.731	-.3273576	.2297988
state_49	.0110171	.1425376	0.08	0.938	-.2684151	.2904493
state_50	.0555716	.1481115	0.38	0.708	-.2347877	.3459309
state_51	-.029108	.1704606	-0.17	0.864	-.3632802	.3050643
_cons	.5838799	.1611747	3.62	0.000	.2679114	.8998484

```
74 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

75 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.1698144	.0488236	3.48	0.001	.0741006	.2655282

```
76 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0628082	.0348338	1.80	0.071	-.00548	.1310965

```
77 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0441979	.0220201	-2.01	0.045	-.0873663	-.0010296

```
78 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.1512041	.014079	-10.74	0.000	-.1788047	-.1236035

```
79 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.2582103	.0187987	-13.74	0.000	-.2950634	-.2213572

```
80 . * COVID-19 Concern, Latent NSC
81 . mi estimate: regress dv_concern c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.r
   > ##c.polengage_scale c.income##c.polengage_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        965
                                                Average RVI       =     0.0112
                                                Largest FMI       =     0.0415
                                                Complete DF       =        946
DF adjustment:   Small sample                   DF:     min       =     850.29
                                                        avg       =     923.19
                                                        max       =     942.89
Model F test:       Equal FMI                   F(  18,  943.7)   =       2.70
Within VCE type:          OLS                   Prob > F          =     0.0002
```

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.3410344	.2394253	1.42	0.155	-.1288337	.8109026
polengage_scale	.2721278	.2842061	0.96	0.339	-.2856298	.8298853
c.latentnsc2#c.polengage_scale	-.5889081	.3060059	-1.92	0.055	-1.189439	.0116233
age	.1872601	.2037141	0.92	0.358	-.2125258	.5870459
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.2445519	.2655462	-0.92	0.357	-.7656824	.2765785
1.male	-.1650795	.0849524	-1.94	0.052	-.3317974	.0016384
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.1564718	.112463	1.39	0.164	-.0642352	.3771788
race						
1	-.1355216	.158605	-0.85	0.393	-.4467892	.175746
2	.2181534	.2067544	1.06	0.292	-.1876464	.6239531
3	.1352213	.1889212	0.72	0.474	-.2355495	.505992
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.1144353	.2234904	0.51	0.609	-.3241758	.5530465
2	-.1953255	.297129	-0.66	0.511	-.7785178	.3878667
3	-.1645444	.2690128	-0.61	0.541	-.6925088	.36342
education	-.011299	.1792973	-0.06	0.950	-.3631737	.3405757
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.1822275	.2418836	0.75	0.451	-.2924722	.6569271
income	-.055832	.1448773	-0.39	0.700	-.3401601	.2284961
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0446706	.1896675	-0.24	0.814	-.4168985	.3275573
5.sample_num	-.0203423	.0263493	-0.77	0.440	-.0720525	.0313679
_cons	.3566723	.2037249	1.75	0.080	-.0431391	.7564836

```
82 . quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

83 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.247932	.1740622	1.42	0.155	-.0936606	.5895247

```
84 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.140898	.1208118	1.17	0.244	-.0961921	.3779881

```
85 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.033864	.0712722	0.48	0.635	-.106006	.1737339

```
86 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0731701	.0418376	-1.75	0.081	-.1552754	.0089352

```
87 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.1802041	.0678782	-2.65	0.008	-.3134134	-.0469949

```
88 . * COVID-19 Restrictions, Authoritarianism
89 . mi estimate: regress dv_restrictions c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scal
   > ation##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      6,870
                                                Average RVI       =     0.0044
                                                Largest FMI       =     0.0853
                                                Complete DF       =       6800
DF adjustment:   Small sample                   DF:     min       =   2,174.22
                                                        avg       =   6,549.62
                                                        max       =   6,797.96
Model F test:       Equal FMI                   F(  69, 6797.4)   =      15.39
Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.1982025	.0342422	5.79	0.000	.1310771	.2653279
polengage_scale	.4185124	.0669678	6.25	0.000	.287234	.5497908
c.auth_scale#c.polengage_scale	-.4771861	.04359	-10.95	0.000	-.5626362	-.3917361
age	.1032363	.0524726	1.97	0.049	.0003607	.2061119
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.2919973	.0716903	-4.07	0.000	-.4325495	-.1514451

1.male	.0668953	.0223694	2.99	0.003	.0230438	.1107469
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.1495827	.0288798	-5.18	0.000	-.2061968	-.0929685
race						
1	-.01038	.0388059	-0.27	0.789	-.0864519	.0656918
2	.006421	.0468799	0.14	0.891	-.0854785	.0983204
3	.0387747	.0466565	0.83	0.406	-.0526866	.130236
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.012385	.0524737	-0.24	0.813	-.1152499	.0904799
2	.1828785	.0657748	2.78	0.005	.0539393	.3118177
3	-.004319	.0658323	-0.07	0.948	-.133371	.1247329
education	.038196	.0430641	0.89	0.375	-.0462251	.1226172
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.0417559	.0554505	0.75	0.451	-.0669472	.1504589
income	.099619	.0453578	2.20	0.028	.0106765	.1885615
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0947405	.0602526	-1.57	0.116	-.2128992	.0234182
sample_num						
4	-.0797173	.0111313	-7.16	0.000	-.1015382	-.0578964
6	.0142547	.0107762	1.32	0.186	-.0068699	.0353794
state_1	.1240173	.0953982	1.30	0.194	-.062993	.3110275
state_2	.0233894	.0790689	0.30	0.767	-.1316105	.1783892
state_3	.0821177	.0803877	1.02	0.307	-.0754673	.2397028
state_4	.0225871	.0763724	0.30	0.767	-.1271267	.1723008
state_5	.1315333	.0744544	1.77	0.077	-.0144207	.2774872
state_6	.0733953	.0776101	0.95	0.344	-.0787448	.2255353
state_7	.0930679	.0819915	1.14	0.256	-.067661	.2537969
state_8	.2732221	.0947601	2.88	0.004	.0874626	.4589816
state_9	.1027545	.0930954	1.10	0.270	-.0797415	.2852506
state_10	.0673181	.0745718	0.90	0.367	-.0788661	.2135022
state_11	.0637246	.0760211	0.84	0.402	-.0853005	.2127498
state_12	.1625155	.0895745	1.81	0.070	-.0130785	.3381096
state_13	.100022	.0807433	1.24	0.215	-.0582601	.2583042
state_14	.0863937	.084935	1.02	0.309	-.0801056	.2528929
state_15	.082832	.075881	1.09	0.275	-.0659185	.2315824
state_16	.0481209	.0774846	0.62	0.535	-.103773	.2000149
state_17	.1072039	.0810532	1.32	0.186	-.0516857	.2660934
state_18	.1145232	.0795437	1.44	0.150	-.0414073	.2704538
state_19	-.0369432	.0807556	-0.46	0.647	-.1952495	.1213631
state_20	.1141451	.0774196	1.47	0.140	-.0376216	.2659118
state_21	.1363348	.0775901	1.76	0.079	-.015766	.2884357
state_22	.1258207	.0865286	1.45	0.146	-.0438024	.2954438
state_23	.0660586	.0759568	0.87	0.385	-.0828406	.2149578
state_24	.0914904	.0788496	1.16	0.246	-.0630794	.2460602
state_25	.0075235	.0771184	0.10	0.922	-.1436528	.1586998
state_26	-.0303459	.0847785	-0.36	0.720	-.1965383	.1358465
state_27	-.0381998	.0900995	-0.42	0.672	-.2148229	.1384234
state_28	.0610487	.076351	0.80	0.424	-.0886231	.2107205
state_29	-.0403775	.1057781	-0.38	0.703	-.2477357	.1669806
state_30	.0610406	.0848956	0.72	0.472	-.1053813	.2274625
state_31	.0071045	.0879304	0.08	0.936	-.1652667	.1794757
state_32	.0601869	.076455	0.79	0.431	-.0896888	.2100625
state_33	-.0577224	.0855445	-0.67	0.500	-.2254163	.1099715
state_34	.0437128	.0807833	0.54	0.588	-.1146478	.2020733

state_35	.1605565	.0748	2.15	0.032	.013925	.307188
state_36	.0879078	.0758124	1.16	0.246	-.0607082	.2365237
state_37	.078821	.0820663	0.96	0.337	-.0820545	.2396966
state_38	.0659131	.079259	0.83	0.406	-.0894593	.2212856
state_39	.0494361	.0750118	0.66	0.510	-.0976105	.1964828
state_40	.1951315	.094535	2.06	0.039	.0098134	.3804497
state_41	.0930807	.0790515	1.18	0.239	-.061885	.2480465
state_42	-.0597531	.0937336	-0.64	0.524	-.2435002	.123994
state_43	.0615299	.0776265	0.79	0.428	-.0906424	.2137022
state_44	.0645122	.0747308	0.86	0.388	-.0819836	.211008
state_45	.0132268	.0822607	0.16	0.872	-.1480299	.1744834
state_46	.0602913	.0767847	0.79	0.432	-.0902308	.2108134
state_47	.1000887	.0954507	1.05	0.294	-.0870244	.2872019
state_48	.0584947	.0767836	0.76	0.446	-.0920251	.2090145
state_49	.0584114	.0772236	0.76	0.449	-.0929709	.2097938
state_50	.1033619	.0836308	1.24	0.217	-.0605806	.2673044
state_51	0	(omitted)				
_cons	.3366418	.0880965	3.82	0.000	.163945	.5093385

```
90 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

91 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.1982025	.0342422	5.79	0.000	.1310771	.2653278

```
92 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0789059	.0241858	3.26	0.001	.0314943	.1263176

```
93 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0403906	.0153261	-2.64	0.008	-.0704345	-.0103467

```
94 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.1596871	.0110605	-14.44	0.000	-.1813693	-.138005

```
95 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.2789837	.0157256	-17.74	0.000	-.3098107	-.2481566

```
96 . * COVID-19 Behaviors, Authoritarianism
97 . mi estimate: regress dv_behavior c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.
   > n##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      4,029
                                                Average RVI       =     0.0040
                                                Largest FMI       =     0.0700
                                                Complete DF       =       3957
DF adjustment:   Small sample                   DF:     min       =   2,111.97
                                                        avg       =   3,818.01
                                                        max       =   3,950.25
Model F test:       Equal FMI                   F(  71, 3954.8)   =      15.08
Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.0367659	.0370423	0.99	0.321	-.035858	.1093898
polengage_scale	.4026752	.0718669	5.60	0.000	.2617753	.5435751
c.auth_scale#c.polengage_scale	-.1207176	.0520662	-2.32	0.020	-.2227969	-.0186383
age	.3338402	.0517109	6.46	0.000	.2324427	.4352376
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.4426272	.0732594	-6.04	0.000	-.5862792	-.2989752
1.male	.0066576	.0220199	0.30	0.762	-.0365173	.0498326
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0470879	.0317922	-1.48	0.139	-.1094226	.0152467
race						
1	-.034041	.0401753	-0.85	0.397	-.1128073	.0447253
2	.0879404	.0479182	1.84	0.067	-.0060063	.1818871
3	.1030468	.0457485	2.25	0.024	.0133536	.19274
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.0248327	.0603273	0.41	0.681	-.0934432	.1431086
2	-.040581	.0744247	-0.55	0.586	-.1864955	.1053335
3	-.1222166	.0708557	-1.72	0.085	-.261134	.0167008
education	.0729715	.043456	1.68	0.093	-.0122295	.1581725
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.0297745	.0625925	-0.48	0.634	-.1524944	.0929453
income	.1033426	.0411838	2.51	0.012	.0225775	.1841077
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0599814	.0581055	-1.03	0.302	-.1739221	.0539593
sample_num						

3	.0122216	.0135802	0.90	0.368	-.0144032	.0388464
4	-.1527008	.0124123	-12.30	0.000	-.177036	-.1283656
5	-.0002833	.0156648	-0.02	0.986	-.0309952	.0304285
state_1	.1200439	.1451191	0.83	0.408	-.1644717	.4045596
state_2	.0116089	.1197683	0.10	0.923	-.2232051	.2464228
state_3	.1019174	.1215034	0.84	0.402	-.1362983	.3401332
state_4	.0629931	.1181404	0.53	0.594	-.1686294	.2946155
state_5	.1362771	.1166251	1.17	0.243	-.0923745	.3649287
state_6	.0200526	.1192456	0.17	0.866	-.2137365	.2538417
state_7	.1256231	.1223928	1.03	0.305	-.1143364	.3655826
state_8	.164833	.1275245	1.29	0.196	-.0851874	.4148535
state_9	.1214752	.1314	0.92	0.355	-.1361433	.3790938
state_10	.0779239	.1167995	0.67	0.505	-.1510695	.3069173
state_11	.08004	.1176083	0.68	0.496	-.1505391	.3106191
state_12	.2583585	.128042	2.02	0.044	.0073234	.5093936
state_13	-.0125625	.1221359	-0.10	0.918	-.2520184	.2268933
state_14	.1450564	.1295566	1.12	0.263	-.1089481	.3990608
state_15	.0613872	.1176885	0.52	0.602	-.1693492	.2921236
state_16	.0349227	.1189422	0.29	0.769	-.1982716	.268117
state_17	.0229152	.12127	0.19	0.850	-.214843	.2606734
state_18	.0962122	.1198931	0.80	0.422	-.1388465	.3312709
state_19	.0540787	.1204065	0.45	0.653	-.1819864	.2901438
state_20	.108403	.1191988	0.91	0.363	-.1252945	.3421004
state_21	.0924213	.1190982	0.78	0.438	-.141079	.3259215
state_22	.1388921	.1297044	1.07	0.284	-.1154022	.3931865
state_23	.0697839	.11792	0.59	0.554	-.1614063	.3009742
state_24	.0575931	.1203137	0.48	0.632	-.1782901	.2934762
state_25	.033491	.1188077	0.28	0.778	-.1994398	.2664217
state_26	.0495724	.1238265	0.40	0.689	-.193198	.2923428
state_27	-.021946	.1317358	-0.17	0.868	-.2802232	.2363311
state_28	.0729592	.1178082	0.62	0.536	-.1580119	.3039302
state_29	-.0877901	.1354494	-0.65	0.517	-.3533478	.1777676
state_30	.0622156	.1257897	0.49	0.621	-.1844036	.3088347
state_31	-.0571627	.1296529	-0.44	0.659	-.3113559	.1970306
state_32	.1078641	.1180417	0.91	0.361	-.1235647	.3392929
state_33	.0780407	.1308409	0.60	0.551	-.1784825	.334564
state_34	.0761606	.1207182	0.63	0.528	-.1605156	.3128369
state_35	.1339861	.1167162	1.15	0.251	-.0948439	.3628162
state_36	.0635035	.1175919	0.54	0.589	-.1670435	.2940505
state_37	.0374509	.1207098	0.31	0.756	-.1992089	.2741108
state_38	.1151749	.1202452	0.96	0.338	-.1205741	.3509238
state_39	.0677143	.1172479	0.58	0.564	-.1621582	.2975868
state_40	.0537204	.1315207	0.41	0.683	-.2041347	.3115755
state_41	.0769336	.119522	0.64	0.520	-.1573976	.3112647
state_42	-.245201	.1555379	-1.58	0.115	-.5501433	.0597413
state_43	.0195644	.1193358	0.16	0.870	-.2144017	.2535304
state_44	.0758343	.116749	0.65	0.516	-.1530601	.3047287
state_45	.0392792	.1234483	0.32	0.750	-.2027495	.281308
state_46	.0737052	.1184957	0.62	0.534	-.1586136	.306024
state_47	.0535106	.1372468	0.39	0.697	-.215571	.3225923
state_48	.0961175	.1189052	0.81	0.419	-.1370043	.3292394
state_49	.0723407	.1198802	0.60	0.546	-.1626926	.307374
state_50	.0926873	.1235224	0.75	0.453	-.1494868	.3348615
state_51	-.0291579	.1371451	-0.21	0.832	-.2980401	.2397243
_cons	.45089	.1260296	3.58	0.000	.2038001	.6979798

```
 98 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

 99 .     lincom _b[6._at]-_b[1._at]

     ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0367659	.0370423	0.99	0.321	-.0358578	.1093897

```
100 .     lincom _b[7._at]-_b[2._at]

     ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0065865	.0253093	0.26	0.795	-.043034	.056207

```
101 .     lincom _b[8._at]-_b[3._at]

     ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0235929	.0157433	-1.50	0.134	-.0544586	.0072728

```
102 .     lincom _b[9._at]-_b[4._at]

     ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0537723	.0139285	-3.86	0.000	-.08108	-.0264646

```
103 .     lincom _b[10._at]-_b[5._at]

     ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0839517	.0218864	-3.84	0.000	-.1268615	-.041042

```
104 . * COVID-19 Behaviors, Latent NSC
105 . mi estimate: regress dv_behavior c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.
    > n##c.polengage_scale c.income##c.polengage_scale i.sample_num

    Multiple-imputation estimates                   Imputations       =         25
    Linear regression                               Number of obs     =        970
                                                    Average RVI       =     0.0100
                                                    Largest FMI       =     0.0267
                                                    Complete DF       =        951
    DF adjustment:   Small sample                   DF:     min       =     899.21
                                                            avg       =     935.95
                                                            max       =     948.39
    Model F test:       Equal FMI                   F(  18,  948.7)   =       4.11
    Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.0666241	.1464614	0.45	0.649	-.2208018	.3540501
polengage_scale	.2590463	.1818714	1.42	0.155	-.0978742	.6159668
c.latentnsc2#c.polengage_scale	-.234829	.1881441	-1.25	0.212	-.6040558	.1343978
age	.0901255	.1246461	0.72	0.470	-.1544888	.3347399
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.1725505	.1629775	-1.06	0.290	-.4923898	.1472888
1.male	-.0707424	.0531278	-1.33	0.183	-.1750042	.0335193
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.0172544	.0704212	0.25	0.806	-.1209454	.1554543
race						
1	-.047128	.1004187	-0.47	0.639	-.2441984	.1499423
2	-.0873595	.1275969	-0.68	0.494	-.3377691	.16305
3	-.0865041	.1229713	-0.70	0.482	-.3278407	.1548324
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.0058429	.1421805	-0.04	0.967	-.2848715	.2731856
2	.1136217	.1851794	0.61	0.540	-.2497945	.4770379
3	.0823609	.1745379	0.47	0.637	-.2601799	.4249018
education	.018429	.1131457	0.16	0.871	-.2036183	.2404763
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.0545586	.1532464	0.36	0.722	-.2461875	.3553048
income	.1224582	.0905643	1.35	0.177	-.0552838	.3002002
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0455863	.1195726	-0.38	0.703	-.2802562	.1890836
5.sample_num	-.0241642	.0163947	-1.47	0.141	-.0563385	.0080102
_cons	.7190468	.1297902	5.54	0.000	.464336	.9737575

```
106 . quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

107 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0484357	.1064775	0.45	0.649	-.1605222	.2573937

```
108 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.0057556**	**.0737614**	**0.08**	**0.938**	**-.1389983**	**.1505094**

```
109 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**-.0369246**	**.0433895**	**-0.85**	**0.395**	**-.1220749**	**.0482256**

```
110 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**-.0796048**	**.0257523**	**-3.09**	**0.002**	**-.1301427**	**-.0290669**

```
111 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**-.122285**	**.0422177**	**-2.90**	**0.004**	**-.2051355**	**-.0394344**

```
112 .
113 . * Standardized Models for d
114 . mi estimate: regress z_dv_concern c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
    >  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num  state_1-state_51

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      5,484
                                                Average RVI       =     0.0037
                                                Largest FMI       =     0.1110
                                                Complete DF       =       5412
DF adjustment:   Small sample                   DF:     min       =   1,402.93
                                                        avg       =   5,221.73
                                                        max       =   5,406.85
Model F test:       Equal FMI                   F(  71, 5409.7)   =       8.30
Within VCE type:          OLS                   Prob > F          =     0.0000
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1288974	.0144133	-8.94	0.000	-.1571533	-.1006415
z_polengage_scale	.0529836	.0706384	0.75	0.453	-.0854963	.1914635
c.z_auth_scale#c.z_polengage_scale	-.1059967	.0147232	-7.20	0.000	-.13486	-.0771334
age	.0422055	.0707845	0.60	0.551	-.0965606	.1809715
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.2195633	.0737417	-2.98	0.003	-.3641267	-.0749999
1.male	-.1590707	.0289746	-5.49	0.000	-.2158726	-.1022687
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.0618112	.0303392	-2.04	0.042	-.1212883	-.0023341
race						
1	-.1881466	.05789	-3.25	0.001	-.3016345	-.0746587
2	.1987479	.0711719	2.79	0.005	.0592219	.3382738
3	.0051046	.0744291	0.07	0.945	-.1408068	.1510159
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0923227	.0577384	1.60	0.110	-.020868	.2055135
2	.0701478	.0686927	1.02	0.307	-.0645178	.2048134
3	.0264831	.0736845	0.36	0.719	-.1179684	.1709347
education	.0688295	.0539728	1.28	0.202	-.0369806	.1746396
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0996942	.05437	1.83	0.067	-.0068935	.2062818
income	-.2035268	.0659163	-3.09	0.002	-.3328319	-.0742217
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	.0015558	.0583848	0.03	0.979	-.1129251	.1160367
sample_num						
2	-.4664103	.0639281	-7.30	0.000	-.5917389	-.3410816
5	-.552779	.0576996	-9.58	0.000	-.665894	-.4396641
6	-.2286309	.0447661	-5.11	0.000	-.3163937	-.140868
state_1	.016512	.4878308	0.03	0.973	-.9398356	.9728595
state_2	-.1029886	.4455278	-0.23	0.817	-.976406	.7704287
state_3	.2232482	.4474741	0.50	0.618	-.6539844	1.100481
state_4	-.2547854	.4339538	-0.59	0.557	-1.105513	.5959423
state_5	.0548061	.4282113	0.13	0.898	-.784664	.8942761
state_6	.025127	.4382939	0.06	0.954	-.8341085	.8843626
state_7	.0851541	.4517825	0.19	0.851	-.8005247	.970833
state_8	.5200265	.5202917	1.00	0.318	-.4999571	1.54001
state_9	.2534801	.4729826	0.54	0.592	-.6737591	1.180719
state_10	.022424	.4285428	0.05	0.958	-.817696	.8625439
state_11	-.0012524	.4317666	-0.00	0.998	-.8476924	.8451876
state_12	.2235539	.4816003	0.46	0.643	-.7205794	1.167687
state_13	.1532342	.4399469	0.35	0.728	-.7092423	1.015711
state_14	-.0338892	.4507765	-0.08	0.940	-.9175957	.8498173
state_15	.1048642	.4311142	0.24	0.808	-.7402967	.950025
state_16	-.0245844	.4363272	-0.06	0.955	-.8799647	.830796
state_17	.0903309	.4443526	0.20	0.839	-.7807823	.9614441
state_18	.1894995	.4431431	0.43	0.669	-.6792428	1.058242
state_19	-.1140272	.4508259	-0.25	0.800	-.9978309	.7697766
state_20	.0486416	.4368377	0.11	0.911	-.8077394	.9050226
state_21	.2922445	.4362157	0.67	0.503	-.5629172	1.147406

state_22	.2458594	.4606557	0.53	0.594	-.6572142	1.148933
state_23	-.0973335	.4316267	-0.23	0.822	-.9434991	.7488322
state_24	-.0732644	.4396339	-0.17	0.868	-.9351274	.7885985
state_25	-.0752477	.4360905	-0.17	0.863	-.9301642	.7796688
state_26	.1642978	.4515443	0.36	0.716	-.7209138	1.049509
state_27	-.1484153	.477867	-0.31	0.756	-1.08523	.7883995
state_28	-.0036848	.4329666	-0.01	0.993	-.8524772	.8451077
state_29	-.5448718	.5991074	-0.91	0.363	-1.719365	.6296218
state_30	-.0048529	.4643601	-0.01	0.992	-.9151886	.9054827
state_31	.0364696	.4644939	0.08	0.937	-.8741285	.9470677
state_32	.0798894	.4334373	0.18	0.854	-.7698255	.9296044
state_33	-.172503	.4572193	-0.38	0.706	-1.06884	.7238338
state_34	.0014553	.4467537	0.00	0.997	-.8743651	.8772756
state_35	.1740861	.4295023	0.41	0.685	-.6679148	1.016087
state_36	.136276	.4316399	0.32	0.752	-.7099154	.9824674
state_37	.2964637	.4528901	0.65	0.513	-.5913867	1.184314
state_38	-.1519864	.4412211	-0.34	0.731	-1.016961	.7129881
state_39	.0832983	.4299506	0.19	0.846	-.7595816	.9261781
state_40	.3123403	.4870922	0.64	0.521	-.6425594	1.26724
state_41	.0252984	.4419353	0.06	0.954	-.8410762	.891673
state_42	.1805261	.4735323	0.38	0.703	-.7477909	1.108843
state_43	.0047964	.4370359	0.01	0.991	-.8519733	.8615662
state_44	-.0299898	.4289037	-0.07	0.944	-.8708172	.8108377
state_45	-.0811312	.4471199	-0.18	0.856	-.9576697	.7954073
state_46	.0196604	.434427	0.05	0.964	-.8319947	.8713156
state_47	.3265047	.5002002	0.65	0.514	-.6540913	1.307101
state_48	-.1488706	.4336835	-0.34	0.731	-.9990682	.701327
state_49	.0336234	.4350129	0.08	0.938	-.8191805	.8864274
state_50	.1696	.4520241	0.38	0.708	-.7165526	1.055753
state_51	-.0888351	.5202316	-0.17	0.864	-1.108701	.9310309
_cons	.4285897	.4347809	0.99	0.324	-.4237609	1.28094

```
115 . mi estimate: regress z_dv_concern c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
>  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates                Imputations       =         25
Linear regression                            Number of obs     =        965
                                             Average RVI       =     0.0112
                                             Largest FMI       =     0.0415
                                             Complete DF       =        946
DF adjustment:   Small sample                DF:     min       =     850.29
                                                     avg       =     926.16
                                                     max       =     943.25
Model F test:       Equal FMI                F(  18,  943.7)   =       2.70
Within VCE type:          OLS                Prob > F          =     0.0002
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.045829	.0370731	-1.24	0.217	-.1185843	.0269262
z_polengage_scale	.0029868	.1936837	0.02	0.988	-.3771217	.3830952
c.z_latentnsc2#c.z_polengage_scale	-.0923728	.0479984	-1.92	0.055	-.1865689	.0018232
age	.045498	.1692745	0.27	0.788	-.2867003	.3776963
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1890517	.2052814	-0.92	0.357	-.5919134	.21381
1.male	-.1672555	.070208	-2.38	0.017	-.3050376	-.0294734
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.120961	.0869399	1.39	0.164	-.0496573	.2915793
race						

1	-.1674628	.1474239	-1.14	0.256	-.4567855	.1218599
2	.2456624	.1850288	1.33	0.185	-.117476	.6088007
3	.0587673	.1791187	0.33	0.743	-.2927629	.4102975
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0884646	.17277	0.51	0.609	-.2506052	.4275345
2	-.1509971	.2296966	-0.66	0.511	-.6018358	.2998417
3	-.1272016	.2079613	-0.61	0.541	-.5353463	.2809431
education	.3574676	.155128	2.30	0.021	.0530287	.6619065
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.1408716	.186989	0.75	0.451	-.2260966	.5078398
income	-.2664761	.131534	-2.03	0.043	-.5246143	-.008338
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0345328	.1466231	-0.24	0.814	-.3222848	.2532193
5.sample_num	-.0620829	.0804159	-0.77	0.440	-.2198981	.0957323
_cons	-.1848749	.1799615	-1.03	0.305	-.5380511	.1683013

116 . mi estimate: regress z_dv_restrictions c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_pole
> scale c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 6,870
Average RVI = 0.0044
Largest FMI = 0.0927
Complete DF = 6800
DF adjustment: **Small sample** — DF: min = 1,942.29
avg = 6,579.18
max = 6,797.96
Model F test: **Equal FMI** — F(69, 6797.4) = 15.39
Within VCE type: **OLS** — Prob > F = 0.0000

z_dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1452909	.011767	-12.35	0.000	-.168358	-.1222238
z_polengage_scale	.1527916	.0520009	2.94	0.003	.050853	.2547302
c.z_auth_scale#c.z_polengage_scale	-.1271633	.0116161	-10.95	0.000	-.1499345	-.1043921
age	-.3368018	.0602837	-5.59	0.000	-.4549775	-.2186262
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.2429063	.0596376	-4.07	0.000	-.3598287	-.1259839
1.male	-.1265237	.0244029	-5.18	0.000	-.1743618	-.0786856
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.1244346	.0240245	-5.18	0.000	-.1715307	-.0773386
race						
1	-.0627554	.0464051	-1.35	0.176	-.153724	.0282131
2	.4443708	.0593011	7.49	0.000	.3281221	.5606195
3	.1173453	.0589651	1.99	0.047	.0017553	.2329353
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	-.0103028	.0436517	-0.24	0.813	-.0958739	.0752683

2	.1521327	.0547166	2.78	0.005	.0448709	.2593945
3	-.0035929	.0547645	-0.07	0.948	-.1109485	.1037626
education	.222088	.0466619	4.76	0.000	.1306142	.3135618
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0347358	.0461281	0.75	0.451	-.0556919	.1251636
income	.1078813	.0510345	2.11	0.035	.0077932	.2079694
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0788126	.0501228	-1.57	0.116	-.1771063	.0194811
sample_num						
4	-.2618038	.036557	-7.16	0.000	-.3334669	-.1901407
6	.0468147	.0353906	1.32	0.186	-.0225619	.1161914
state_1	.4072915	.3133021	1.30	0.194	-.2068787	1.021462
state_2	.0768143	.2596744	0.30	0.767	-.4322289	.5858574
state_3	.2696871	.2640055	1.02	0.307	-.2478462	.7872204
state_4	.0741794	.2508185	0.30	0.767	-.4175033	.5658622
state_5	.4319752	.2445196	1.77	0.077	-.0473598	.9113103
state_6	.2410412	.2548834	0.95	0.344	-.2586099	.7406924
state_7	.3056492	.2692725	1.14	0.256	-.2222092	.8335075
state_8	.8973029	.3112066	2.88	0.004	.2872405	1.507365
state_9	.3374615	.3057393	1.10	0.270	-.2618832	.9368063
state_10	.2210828	.2449053	0.90	0.367	-.2590081	.7011738
state_11	.2092814	.2496649	0.84	0.402	-.2801398	.6987027
state_12	.5337257	.2941762	1.81	0.070	-.0429517	1.110403
state_13	.3284876	.2651732	1.24	0.215	-.1913349	.8483102
state_14	.28373	.2789396	1.02	0.309	-.2630789	.8305389
state_15	.2720328	.2492047	1.09	0.275	-.2164864	.7605519
state_16	.1580365	.2544711	0.62	0.535	-.3408064	.6568794
state_17	.3520738	.2661909	1.32	0.186	-.1697437	.8738912
state_18	.3761117	.2612336	1.44	0.150	-.1359879	.8882113
state_19	-.121327	.2652138	-0.46	0.647	-.641229	.398575
state_20	.3748697	.2542578	1.47	0.140	-.1235552	.8732947
state_21	.4477444	.2548176	1.76	0.079	-.0517779	.9472666
state_22	.4132143	.284173	1.45	0.146	-.1438536	.9702823
state_23	.2169465	.2494538	0.87	0.385	-.2720611	.7059541
state_24	.3004684	.258954	1.16	0.246	-.2071624	.8080993
state_25	.0247084	.2532687	0.10	0.922	-.4717776	.5211944
state_26	-.0996606	.2784255	-0.36	0.720	-.6454617	.4461405
state_27	-.1254538	.2959003	-0.42	0.672	-.7055112	.4546035
state_28	.2004933	.2507482	0.80	0.424	-.2910517	.6920382
state_29	-.1326059	.3473913	-0.38	0.703	-.8136017	.5483898
state_30	.2004666	.27881	0.72	0.472	-.3460884	.7470215
state_31	.0233323	.2887769	0.08	0.936	-.5427609	.5894255
state_32	.1976629	.2510897	0.79	0.431	-.2945515	.6898773
state_33	-.1895692	.280941	-0.67	0.500	-.7403015	.3611632
state_34	.1435594	.2653047	0.54	0.588	-.3765208	.6636396
state_35	.5272921	.2456547	2.15	0.032	.0457319	1.008852
state_36	.2887025	.2489793	1.16	0.246	-.1993749	.7767799
state_37	.2588602	.2695181	0.96	0.337	-.2694796	.7872
state_38	.2164688	.2602985	0.83	0.406	-.2937978	.7267354
state_39	.1623558	.2463502	0.66	0.510	-.3205676	.6452793
state_40	.6408417	.3104673	2.06	0.039	.0322285	1.249455
state_41	.3056912	.2596173	1.18	0.239	-.2032399	.8146224
state_42	-.1962383	.3078353	-0.64	0.524	-.7996918	.4072153
state_43	.2020736	.2549374	0.79	0.428	-.2976835	.7018306
state_44	.211868	.2454274	0.86	0.388	-.2692465	.6929825
state_45	.0434388	.2701565	0.16	0.872	-.4861526	.5730302
state_46	.1980058	.2521727	0.79	0.432	-.2963317	.6923434
state_47	.3287067	.3134745	1.05	0.294	-.2858015	.9432148
state_48	.1921053	.2521689	0.76	0.446	-.3022246	.6864353
state_49	.1918321	.2536139	0.76	0.449	-.3053306	.6889947
state_50	.3394563	.2746562	1.24	0.217	-.1989558	.8778684

state_51	0	(omitted)				
_cons	-.1555637	.2500558	-0.62	0.534	-.6457515	.334624

```
117 . mi estimate: regress z_dv_behavior c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
> e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num  state_1-state_51

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =      4,029
                                              Average RVI       =     0.0040
                                              Largest FMI       =     0.0580
                                              Complete DF       =       3957
DF adjustment:   Small sample                 DF:     min       =   2,456.86
                                                      avg       =   3,869.56
                                                      max       =   3,950.96
Model F test:       Equal FMI                 F(  71, 3954.8)   =      15.08
Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.0611614	.0169203	-3.61	0.000	-.0943348	-.027988
z_polengage_scale	.3436187	.0673692	5.10	0.000	.2115366	.4757007
c.z_auth_scale#c.z_polengage_scale	-.0387107	.0166962	-2.32	0.020	-.0714447	-.0059768
age	.0865129	.0732158	1.18	0.237	-.0570351	.2300609
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.4430828	.0733348	-6.04	0.000	-.5868827	-.2992829
1.male	-.1048385	.0316475	-3.31	0.001	-.166886	-.042791
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.0471364	.031825	-1.48	0.139	-.1095352	.0152624
race						
1	-.0653641	.0618411	-1.06	0.291	-.1866079	.0558797
2	.2345092	.0792877	2.96	0.003	.0790604	.389958
3	.0668377	.0747687	0.89	0.371	-.0797515	.2134268
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0248582	.0603894	0.41	0.681	-.0935394	.1432559
2	-.0406227	.0745013	-0.55	0.586	-.1866874	.1054419
3	-.1223424	.0709286	-1.72	0.085	-.2614027	.016718
education	.2054511	.0640001	3.21	0.001	.0799746	.3309275
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	-.0298052	.0626569	-0.48	0.634	-.1526513	.093041
income	.2413437	.057582	4.19	0.000	.1284487	.3542386
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0600431	.0581653	-1.03	0.302	-.1741011	.0540149
sample_num						
3	.048299	.053668	0.90	0.368	-.0569207	.1535187
4	-.6034641	.0490528	-12.30	0.000	-.6996352	-.5072929
5	-.0011198	.0619062	-0.02	0.986	-.122491	.1202514
state_1	.474406	.5735015	0.83	0.408	-.6499819	1.598794
state_2	.0458776	.4733168	0.10	0.923	-.8820924	.9738475
state_3	.4027713	.4801739	0.84	0.402	-.5386425	1.344185

state_4	.2489446	.4668835	0.53	0.594	-.6664126	1.164302
state_5	.5385585	.4608952	1.17	0.243	-.3650582	1.442175
state_6	.0792465	.471251	0.17	0.866	-.8446734	1.003166
state_7	.4964547	.4836888	1.03	0.305	-.4518502	1.44476
state_8	.6514098	.5039688	1.29	0.196	-.3366553	1.639475
state_9	.4800624	.5192845	0.92	0.355	-.5380299	1.498155
state_10	.3079503	.4615841	0.67	0.505	-.5970171	1.212918
state_11	.3163131	.4647806	0.68	0.496	-.5949211	1.227547
state_12	1.021017	.506014	2.02	0.044	.0289417	2.013091
state_13	-.0496464	.4826736	-0.10	0.918	-.9959608	.8966681
state_14	.5732536	.5119995	1.12	0.263	-.4305559	1.577063
state_15	.2425983	.4650975	0.52	0.602	-.6692574	1.154454
state_16	.1380123	.470052	0.29	0.769	-.7835568	1.059582
state_17	.0905594	.4792515	0.19	0.850	-.849046	1.030165
state_18	.3802245	.4738101	0.80	0.422	-.5487126	1.309162
state_19	.2137155	.4758387	0.45	0.653	-.7191989	1.14663
state_20	.4284017	.4710662	0.91	0.363	-.495156	1.351959
state_21	.365243	.4706688	0.78	0.438	-.5575353	1.288021
state_22	.5488929	.5125838	1.07	0.284	-.4560624	1.553848
state_23	.2757817	.4660124	0.59	0.554	-.6378677	1.189431
state_24	.2276042	.4754722	0.48	0.632	-.7045912	1.159799
state_25	.1323541	.4695207	0.28	0.778	-.7881735	1.052882
state_26	.195907	.4893547	0.40	0.689	-.7635064	1.15532
state_27	-.0867293	.5206117	-0.17	0.868	-1.107424	.9339656
state_28	.28833	.4655705	0.62	0.536	-.6244529	1.201113
state_29	-.3469409	.5352876	-0.65	0.517	-1.396408	.7025264
state_30	.2458721	.4971128	0.49	0.621	-.7287513	1.220496
state_31	-.2259033	.51238	-0.44	0.659	-1.230459	.7786525
state_32	.426272	.4664933	0.91	0.361	-.4883201	1.340864
state_33	.308412	.5170749	0.60	0.551	-.7053517	1.322176
state_34	.3009821	.4770707	0.63	0.528	-.6343477	1.236312
state_35	.5295048	.4612549	1.15	0.251	-.3748173	1.433827
state_36	.2509619	.4647157	0.54	0.589	-.6601453	1.162069
state_37	.1480037	.4770375	0.31	0.756	-.7872612	1.083268
state_38	.4551638	.4752014	0.96	0.338	-.4765012	1.386829
state_39	.2676028	.4633561	0.58	0.564	-.6408389	1.176044
state_40	.2122995	.5197614	0.41	0.683	-.8067277	1.231327
state_41	.3040367	.4723436	0.64	0.520	-.6220253	1.230099
state_42	-.9690191	.6146758	-1.58	0.115	-2.174132	.2360937
state_43	.0773172	.4716075	0.16	0.870	-.8473019	1.001936
state_44	.2996923	.4613846	0.65	0.516	-.604884	1.204269
state_45	.1552291	.4878598	0.32	0.750	-.8012532	1.111711
state_46	.2912784	.4682874	0.62	0.534	-.626831	1.209388
state_47	.2114707	.5423908	0.39	0.697	-.8519229	1.274864
state_48	.3798505	.4699061	0.81	0.419	-.5414324	1.301133
state_49	.2858859	.4737588	0.60	0.546	-.6429508	1.214723
state_50	.3662946	.4881529	0.75	0.453	-.5907624	1.323352
state_51	-.1152303	.5419889	-0.21	0.832	-1.177836	.9473753
_cons	-.2147736	.4659779	-0.46	0.645	-1.128355	.698808

```
118 . mi estimate: regress z_dv_behavior c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
> e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        970
                                                Average RVI       =     0.0100
                                                Largest FMI       =     0.0224
                                                Complete DF       =        951
DF adjustment:   Small sample                   DF:     min       =     910.40
                                                        avg       =     933.13
                                                        max       =     948.29
Model F test:       Equal FMI                   F(  18,  948.7)   =       4.11
Within VCE type:          OLS                   Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.0792841	.0294006	-2.70	0.007	-.1369819	-.0215864
z_polengage_scale	.1522316	.1624136	0.94	0.349	-.1665041	.4709673
c.z_latentnsc2#c.z_polengage_scale	-.0476964	.0382142	-1.25	0.212	-.1226906	.0272977
age	-.124416	.1343222	-0.93	0.355	-.3880218	.1391899
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1727281	.1631452	-1.06	0.290	-.4928966	.1474404
1.male	-.2315128	.0565879	-4.09	0.000	-.3425658	-.1204597
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.0172722	.0704937	0.25	0.806	-.1210699	.1556143
race						
1	-.2025207	.1174573	-1.72	0.085	-.4330358	.0279944
2	-.0287809	.1472701	-0.20	0.845	-.3178022	.2602403
3	-.1124678	.14412	-0.78	0.435	-.3953133	.1703776
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	-.0058489	.1423268	-0.04	0.967	-.2851647	.2734668
2	.1137387	.18537	0.61	0.540	-.2500516	.4775289
3	.0824457	.1747175	0.47	0.637	-.2604477	.4253391
education	.2247866	.1271918	1.77	0.078	-.0248285	.4744017
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0546148	.1534042	0.36	0.722	-.2464409	.3556705
income	.3569805	.1071197	3.33	0.001	.1467532	.5672078
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0456332	.1196956	-0.38	0.703	-.2805446	.1892783
5.sample_num	-.0954953	.0647909	-1.47	0.141	-.2226464	.0316557
_cons	.3608952	.1436351	2.51	0.012	.0790087	.6427817

```
119 .
120 .
    end of do-file

121 .
```

```
10 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000001.tmp"

11 . *Note that the firthlogits take forever when combined with 25 MIs, quickest way to reproduce 99% similar results is to
   > ns.
12 .
13 . local vars dv_distancing dv_masking

14 . foreach var of local vars {
     2. mi estimate: regress `var' c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.rac
   > c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51
     3. quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post
     4.     lincom _b[6._at]-_b[1._at]
     5.     lincom _b[7._at]-_b[2._at]
     6.     lincom _b[8._at]-_b[3._at]
     7.     lincom _b[9._at]-_b[4._at]
     8.     lincom _b[10._at]-_b[5._at]
     9. }

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      4,023
                                                Average RVI       =     0.0027
                                                Largest FMI       =     0.0299
                                                Complete DF       =       3951
DF adjustment:   Small sample                   DF:     min       =   3,355.95
                                                        avg       =   3,880.65
                                                        max       =   3,944.38
Model F test:       Equal FMI                   F(  71, 3948.9)   =       6.25
Within VCE type:          OLS                   Prob > F          =     0.0000

------------------------------------------------------------------------------
        dv_distancing | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
----------------------+-------------------------------------------------------
           auth_scale |  -.0220258   .0463732    -0.47   0.635    -.1129436     .068892
     polengage_scale |   .3155937   .0899724     3.51   0.000     .1391965    .4919909
                     |
c.auth_scale#c.polen~e |   .0087522   .0651741     0.13   0.893     -.119026    .1365303
                     |
                 age |   .3327543   .0638195     5.21   0.000     .2076281    .4578806
     polengage_scale |          0  (omitted)
                     |
c.age#c.polengage_s~e |  -.4994617   .0903927    -5.53   0.000    -.6766888   -.3222347
                     |
              1.male |  -.0074478   .0273029    -0.27   0.785     -.060978    .0460825
     polengage_scale |          0  (omitted)
                     |
male#c.polengage_sc~e |
                   1 |  -.0538389    .039484    -1.36   0.173    -.1312513    .0235734
                     |
                race |
                   1 |   .0304657   .0502802     0.61   0.545    -.0681121    .1290434
                   2 |   .1551513   .0600075     2.59   0.010     .0375026    .2728001
                   3 |   .1407914   .0572492     2.46   0.014     .0285504    .2530324
                     |
     polengage_scale |          0  (omitted)
                     |
race#c.polengage_sc~e |
                   1 |  -.0774408   .0755772    -1.02   0.306    -.2256153    .0707338
                   2 |  -.1086274   .0932703    -1.16   0.244    -.2914901    .0742354
                   3 |  -.1509935   .0888027    -1.70   0.089    -.3250978    .0231107
```

education	.0104971	.0542	0.19	0.846	-.0957666	.1167608
1.polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.0072902	.0780848	0.09	0.926	-.1458017	.160382
income	.001383	.0505324	0.03	0.978	-.0976943	.1004604
1.polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.0653295	.0715348	0.91	0.361	-.0749245	.2055836
sample_num						
3	-.0343569	.0170218	-2.02	0.044	-.0677293	-.0009845
4	-.1393319	.0155608	-8.95	0.000	-.1698399	-.108824
5	.0093178	.0196719	0.47	0.636	-.0292503	.0478859
state_1	.0944438	.1817931	0.52	0.603	-.261974	.4508616
state_2	.0753655	.1500691	0.50	0.616	-.2188558	.3695868
state_3	.1328086	.152233	0.87	0.383	-.1656551	.4312723
state_4	.1092077	.1480111	0.74	0.461	-.1809787	.399394
state_5	.1657809	.1461435	1.13	0.257	-.1207441	.4523058
state_6	.0582834	.1494057	0.39	0.696	-.2346372	.3512039
state_7	.1366365	.1533405	0.89	0.373	-.1639985	.4372715
state_8	.2218788	.1597612	1.39	0.165	-.0913443	.5351018
state_9	.1610536	.1646094	0.98	0.328	-.1616747	.483782
state_10	.1175722	.14635	0.80	0.422	-.1693575	.404502
state_11	.1148607	.1473551	0.78	0.436	-.1740396	.403761
state_12	.3048066	.1604151	1.90	0.057	-.0096985	.6193116
state_13	-.0077295	.1532364	-0.05	0.960	-.3081604	.2927014
state_14	.1144283	.162322	0.70	0.481	-.2038155	.432672
state_15	.1242136	.1474541	0.84	0.400	-.1648809	.413308
state_16	.076538	.1490257	0.51	0.608	-.2156375	.3687135
state_17	.0898099	.1519478	0.59	0.555	-.2080947	.3877145
state_18	.1788176	.1502175	1.19	0.234	-.1156946	.4733299
state_19	.0774768	.150885	0.51	0.608	-.2183441	.3732977
state_20	.1221715	.1494118	0.82	0.414	-.1707611	.4151041
state_21	.1442883	.1492043	0.97	0.334	-.1482374	.4368139
state_22	.2490593	.162486	1.53	0.125	-.0695061	.5676246
state_23	.1177177	.1477417	0.80	0.426	-.1719406	.407376
state_24	.0758962	.1507608	0.50	0.615	-.219681	.3714734
state_25	.1142374	.1488578	0.77	0.443	-.1776089	.4060838
state_26	.0727696	.155144	0.47	0.639	-.2314013	.3769405
state_27	.0216427	.1650422	0.13	0.896	-.3019342	.3452196
state_28	.1290248	.1476097	0.87	0.382	-.1603746	.4184242
state_29	.0373057	.169679	0.22	0.826	-.2953618	.3699732
state_30	.0928362	.1575956	0.59	0.556	-.2161411	.4018136
state_31	-.0036202	.1624243	-0.02	0.982	-.3220644	.3148241
state_32	.1423881	.147894	0.96	0.336	-.1475686	.4323449
state_33	.0592876	.1637673	0.36	0.717	-.2617903	.3803654
state_34	.161144	.1512548	1.07	0.287	-.1354019	.4576899
state_35	.1708957	.1462458	1.17	0.243	-.1158298	.4576212
state_36	.1377487	.1473443	0.93	0.350	-.1511305	.4266279
state_37	.0852022	.1512181	0.56	0.573	-.2112717	.381676
state_38	.1670524	.1506479	1.11	0.268	-.1283035	.4624083
state_39	.0713353	.1469032	0.49	0.627	-.216679	.3593497
state_40	.1106093	.1647649	0.67	0.502	-.2124238	.4336424
state_41	.138384	.149839	0.92	0.356	-.1553861	.4321541
state_42	-.149297	.1947976	-0.77	0.443	-.531211	.232617
state_43	.1101269	.1495282	0.74	0.461	-.183034	.4032877
state_44	.1050217	.1462862	0.72	0.473	-.1817829	.3918263
state_45	.1681789	.1546693	1.09	0.277	-.1350613	.471419
state_46	.0979298	.1484613	0.66	0.510	-.1931391	.3889987
state_47	.1218948	.171925	0.71	0.478	-.215176	.4589657
state_48	.0927759	.1489841	0.62	0.534	-.199318	.3848699
state_49	.0924406	.1501982	0.62	0.538	-.2020336	.3869149
state_50	.1383124	.1547634	0.89	0.372	-.1651123	.4417371
state_51	.0424087	.1717949	0.25	0.805	-.2944071	.3792244

_cons	.5198784	.157871	3.29	0.001	.2103609	.8293959

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0220258	.0463732	-0.47	0.635	-.1129435	.0688919

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0198377	.0316877	-0.63	0.531	-.0819635	.0422881

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0176497	.0197156	-0.90	0.371	-.0563034	.021004

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0154617	.0174429	-0.89	0.375	-.0496596	.0187363

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0132736	.0274001	-0.48	0.628	-.0669933	.040446

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 4,017
Average RVI = 0.0029
Largest FMI = 0.0632
Complete DF = 3945
DF adjustment: Small sample — DF: min = 2,298.30
avg = 3,853.85
max = 3,942.95
Model F test: Equal FMI — F(71, 3942.9) = 26.45
Within VCE type: OLS — Prob > F = 0.0000

dv_masking	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.0325053	.0418969	0.78	0.438	-.0496364	.114647
polengage_scale	.3536143	.0814013	4.34	0.000	.1940215	.5132071
c.auth_scale#c.polengage_scale	-.0963383	.0589013	-1.64	0.102	-.2118183	.0191417
age	.1925143	.0588728	3.27	0.001	.0770657	.3079629
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.364463	.0834031	-4.37	0.000	-.5280163	-.2009098

1.male	.0114782	.024874	0.46	0.645	-.0372918	.0602481
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0520739	.0358879	-1.45	0.147	-.1224375	.0182897
race						
1	-.0246345	.0456044	-0.54	0.589	-.1140451	.0647761
2	.1589664	.0543418	2.93	0.003	.0524256	.2655072
3	.1218568	.0518292	2.35	0.019	.0202422	.2234713
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.0008974	.0684224	-0.01	0.990	-.1350444	.1332495
2	-.0855233	.084313	-1.01	0.310	-.2508244	.0797779
3	-.1609213	.0803235	-2.00	0.045	-.3184011	-.0034416
education	.0452926	.048977	0.92	0.355	-.0507309	.141316
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.0416877	.0705442	-0.59	0.555	-.1799953	.09662
income	.0581695	.0460299	1.26	0.206	-.0320839	.1484229
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0267691	.0651026	-0.41	0.681	-.1544163	.100878
sample_num						
3	-.1026136	.0154012	-6.66	0.000	-.1328087	-.0724186
4	-.3581273	.0140807	-25.43	0.000	-.3857334	-.3305212
5	-.0296628	.017793	-1.67	0.096	-.0645472	.0052216
state_1	.0401488	.1805798	0.22	0.824	-.3138898	.3941873
state_2	-.0599347	.1550163	-0.39	0.699	-.3638543	.2439849
state_3	-.0119081	.156694	-0.08	0.939	-.319117	.2953007
state_4	-.0069474	.1534251	-0.05	0.964	-.3077473	.2938525
state_5	.066492	.1519394	0.44	0.662	-.2313952	.3643792
state_6	-.1116383	.1544952	-0.72	0.470	-.4145363	.1912597
state_7	-.0140829	.1576416	-0.09	0.929	-.3231496	.2949839
state_8	.0647	.1626931	0.40	0.691	-.2542704	.3836705
state_9	.0556448	.1666074	0.33	0.738	-.2710001	.3822896
state_10	-.0249044	.1520971	-0.16	0.870	-.3231008	.2732921
state_11	-.028445	.1528562	-0.19	0.852	-.3281295	.2712396
state_12	.2091192	.1632556	1.28	0.200	-.1109541	.5291925
state_13	-.0946256	.1573166	-0.60	0.548	-.4030552	.213804
state_14	.0353478	.1648125	0.21	0.830	-.2877778	.3584735
state_15	-.0337439	.1529613	-0.22	0.825	-.3336347	.2661469
state_16	-.0604566	.15422	-0.39	0.695	-.362815	.2419019
state_17	-.1023641	.1565247	-0.65	0.513	-.4092412	.204513
state_18	-.0126662	.1552166	-0.08	0.935	-.3169786	.2916462
state_19	-.0312737	.1556297	-0.20	0.841	-.336396	.2738487
state_20	.015424	.1544559	0.10	0.920	-.2873969	.3182449
state_21	-.0054556	.1543352	-0.04	0.972	-.3080398	.2971287
state_22	-.014749	.1649477	-0.09	0.929	-.3381399	.3086419
state_23	-.0477123	.1532056	-0.31	0.755	-.3480819	.2526573
state_24	-.0702133	.1555848	-0.45	0.652	-.3752475	.2348209
state_25	-.0428951	.1540652	-0.28	0.781	-.34495	.2591598
state_26	-.0681238	.1589704	-0.43	0.668	-.3797957	.2435481
state_27	-.1696211	.1669662	-1.02	0.310	-.4969694	.1577271
state_28	-.0023986	.1530597	-0.02	0.987	-.3024822	.297685
state_29	-.2935732	.1707746	-1.72	0.086	-.6283881	.0412417
state_30	-.0146723	.1610526	-0.09	0.927	-.3304265	.3010819
state_31	-.19396	.1648732	-1.18	0.239	-.5172048	.1292848
state_32	.0055493	.1533249	0.04	0.971	-.2950542	.3061528
state_33	-.0074902	.1659144	-0.05	0.964	-.3327763	.3177959
state_34	.0841427	.1560725	0.54	0.590	-.2218477	.3901331

state_35	.0512094	.152007	0.34	0.736	-.2468103	.3492292
state_36	-.0559911	.152871	-0.37	0.714	-.3557048	.2437226
state_37	-.0360039	.1559075	-0.23	0.817	-.3416707	.2696629
state_38	.0569612	.1555259	0.37	0.714	-.2479576	.36188
state_39	-.0290495	.152522	-0.19	0.849	-.3280789	.2699799
state_40	.0056978	.1668343	0.03	0.973	-.3213918	.3327875
state_41	-.0269661	.1547599	-0.17	0.862	-.3303831	.2764509
state_42	-.3650519	.1913794	-1.91	0.057	-.7402637	.01016
state_43	-.0631182	.1545767	-0.41	0.683	-.3661761	.2399396
state_44	-.0253452	.15205	-0.17	0.868	-.3234493	.2727589
state_45	-.0788299	.158697	-0.50	0.619	-.3899658	.232306
state_46	-.0450638	.1538041	-0.29	0.770	-.3466069	.2564793
state_47	-.1020372	.1725789	-0.59	0.554	-.4403896	.2363152
state_48	-.0127915	.1542016	-0.08	0.934	-.3151139	.2895309
state_49	-.1068666	.1551197	-0.69	0.491	-.4109891	.1972558
state_50	-.0177875	.158808	-0.11	0.911	-.3291411	.2935661
state_51	-.2629621	.1724578	-1.52	0.127	-.6010771	.0751528
_cons	.7848671	.1615701	4.86	0.000	.4680982	1.101636

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0325053	.0418969	0.78	0.438	-.0496363	.1146469

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0084207	.0286263	0.29	0.769	-.047703	.0645444

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0156639	.0178114	-0.88	0.379	-.0505843	.0192566

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0397484	.0157703	-2.52	0.012	-.0706671	-.0088298

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.063833	.0247755	-2.58	0.010	-.1124071	-.0152589

```
15 .
16 . replace dv_vaxstatus = 1 if dv_vaxstatus == .5
   (229 real changes made)

17 . mi estimate, cmdok: firthlogit dv_vaxstatus c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polenga
   > cale

   Multiple-imputation estimates                 Imputations       =         25
                                                 Number of obs     =      2,039
                                                 Average RVI       =     0.0126
                                                 Largest FMI       =     0.0602
   DF adjustment:   Large sample                 DF:     min       =   6,686.50
                                                         avg       = 8020488.84
                                                         max       =   4.35e+07
   Model F test:       Equal FMI                 F(  17, 2.3e+06)  =      12.32
   Within VCE type:          OIM                 Prob > F          =     0.0000
```

dv_vaxstatus	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.4972759	.4549682	1.09	0.274	-.3944457	1.388998
polengage_scale	2.645072	.989026	2.67	0.007	.7066155	4.583528
c.auth_scale#c.polengage_scale	-1.610134	.7273866	-2.21	0.027	-3.035785	-.1844818
age	1.89727	.6345268	2.99	0.003	.6534813	3.141059
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-1.573383	1.018216	-1.55	0.122	-3.56922	.4224532
1.male	.2673684	.2699453	0.99	0.322	-.2617685	.7965053
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0952525	.4435277	-0.21	0.830	-.9646139	.7741089
race						
1	-.3652642	.4645922	-0.79	0.432	-1.275848	.5453199
2	-.1674728	.554196	-0.30	0.763	-1.253677	.9187315
3	.506389	.5422814	0.93	0.350	-.5564631	1.569241
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.5713857	.8037542	0.71	0.477	-1.003944	2.146715
2	-.003328	.9808555	-0.00	0.997	-1.92577	1.919114
3	-.5194378	.9566212	-0.54	0.587	-2.394381	1.355505
education	1.002694	.5403846	1.86	0.064	-.056445	2.061834
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.3811864	.8718302	-0.44	0.662	-2.089949	1.327576
income	1.24807	.5295373	2.36	0.018	.2100078	2.286131
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.326728	.8509589	-0.38	0.701	-1.994789	1.341333
_cons	-1.580638	.565373	-2.80	0.005	-2.688749	-.4725267

```
18 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1) race=1) atmeans post expression(exp(predict(xb)
19 .     lincom _b[6._at]-_b[1._at]
 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.1225415	.1110737	1.10	0.270	-.0951589	.3402418

```
20 .     lincom _b[7._at]-_b[2._at]
 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.02198	.0694461	0.32	0.752	-.1141319	.158092

```
21 .     lincom _b[8._at]-_b[3._at]
 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0630171	.0400499	-1.57	0.116	-.1415134	.0154792

```
22 .     lincom _b[9._at]-_b[4._at]
 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1219857	.0397986	-3.07	0.002	-.1999896	-.0439819

```
23 .     lincom _b[10._at]-_b[5._at]
 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1548007	.0531071	-2.91	0.004	-.2588887	-.0507128

```
24 .
25 . local vars dv_travel dv_work_home dv_wash_hands dv_avoid_dine dv_cancel_event dv_sanitize
```

```
26 . foreach var of local vars {
  2. mi estimate, cmdok: firthlogit `var' c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_s
>  i.sample_num
  3. quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1) race=1 sample_num=2) atmeans post expression(
  4.     lincom _b[6._at]-_b[1._at]
  5.     lincom _b[7._at]-_b[2._at]
  6.     lincom _b[8._at]-_b[3._at]
  7.     lincom _b[9._at]-_b[4._at]
  8.     lincom _b[10._at]-_b[5._at]
  9. }
```

Multiple-imputation estimates

Imputations	=	**25**	
Number of obs	=	**958**	
Average RVI	=	**0.0071**	
Largest FMI	=	**0.0283**	
DF adjustment: **Large sample**			
DF: min	=	**30,210.15**	
avg	=	**7638618.91**	
max	=	**5.94e+07**	
Model F test: **Equal FMI**			
F(**18, 7.9e+06**)	=	**3.14**	
Within VCE type: **OIM**			
Prob > F	=	**0.0000**	

dv_travel	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	**-2.889227**	**1.409679**	**-2.05**	**0.040**	**-5.652147**	**-.1263065**
polengage_scale	**-1.508948**	**2.083722**	**-0.72**	**0.469**	**-5.592988**	**2.575092**
c.auth_scale#c.polengage_scale	**3.264154**	**1.843905**	**1.77**	**0.077**	**-.3498334**	**6.878142**
age	**.9152781**	**1.292127**	**0.71**	**0.479**	**-1.617245**	**3.447802**
polengage_scale	**0**	(omitted)				
c.age#c.polengage_scale	**-2.014588**	**1.724415**	**-1.17**	**0.243**	**-5.394379**	**1.365203**
1.male	**-.455234**	**.5608916**	**-0.81**	**0.417**	**-1.554562**	**.6440936**
polengage_scale	**0**	(omitted)				
male#c.polengage_scale						
1	**.0551745**	**.7576812**	**0.07**	**0.942**	**-1.429854**	**1.540203**
race						
1	**-1.419107**	**1.195716**	**-1.19**	**0.235**	**-3.762695**	**.9244808**
2	**-.4651655**	**1.452322**	**-0.32**	**0.749**	**-3.311683**	**2.381352**
3	**-1.53915**	**1.396187**	**-1.10**	**0.270**	**-4.27567**	**1.19737**
polengage_scale	**0**	(omitted)				
race#c.polengage_scale						
1	**1.591332**	**1.686143**	**0.94**	**0.345**	**-1.713476**	**4.89614**
2	**.6268347**	**2.111529**	**0.30**	**0.767**	**-3.511711**	**4.76538**
3	**1.908181**	**1.995206**	**0.96**	**0.339**	**-2.002392**	**5.818754**
education	**-.0045646**	**1.237161**	**-0.00**	**0.997**	**-2.429417**	**2.420287**
polengage_scale	**0**	(omitted)				
c.education#c.polengage_scale	**1.203135**	**1.700863**	**0.71**	**0.479**	**-2.130557**	**4.536827**
income	**2.231536**	**1.071241**	**2.08**	**0.037**	**.1318581**	**4.331213**
polengage_scale	**0**	(omitted)				
c.income#c.polengage_scale	**-1.289969**	**1.423825**	**-0.91**	**0.365**	**-4.080717**	**1.500779**
5.sample_num	**-.3303204**	**.1810378**	**-1.82**	**0.068**	**-.6851481**	**.0245072**
_cons	**2.220303**	**1.483175**	**1.50**	**0.134**	**-.6866826**	**5.127289**

(1) **- 1bn._at + 6._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.6014351**	**.2218836**	**-2.71**	**0.007**	**-1.036319**	**-.1665512**

(1) **- 2._at + 7._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.4326243**	**.1822209**	**-2.37**	**0.018**	**-.7897708**	**-.0754778**

(1) **- 3._at + 8._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.2447656**	**.1086423**	**-2.25**	**0.024**	**-.4577006**	**-.0318306**

(1) **- 4._at + 9._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.0750613**	**.0618747**	**-1.21**	**0.225**	**-.1963334**	**.0462109**

(1) **- 5._at + 10._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**.05362**	**.0841731**	**0.64**	**0.524**	**-.1113562**	**.2185962**

Multiple-imputation estimates — Imputations = **25**
Number of obs = **951**
Average RVI = **0.0108**
Largest FMI = **0.0270**
DF adjustment: **Large sample** — DF: min = **33,097.44**
avg = **2.25e+07**
max = **1.97e+08**
Model F test: **Equal FMI** — F(**18, 3.4e+06**) = **4.64**
Within VCE type: **OIM** — Prob > F = **0.0000**

dv_work_home	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	**.269864**	**1.316484**	**0.20**	**0.838**	**-2.310397**	**2.850125**
polengage_scale	**2.810576**	**1.969356**	**1.43**	**0.154**	**-1.049296**	**6.670448**
c.auth_scale#c.polengage_scale	**-1.5496**	**1.728098**	**-0.90**	**0.370**	**-4.936609**	**1.837409**
age	**.4656674**	**1.266649**	**0.37**	**0.713**	**-2.016919**	**2.948254**
polengage_scale	**0**	(omitted)				
c.age#c.polengage_scale	**-1.73062**	**1.66999**	**-1.04**	**0.300**	**-5.003741**	**1.5425**
1.male	**-.1088667**	**.5306183**	**-0.21**	**0.837**	**-1.14886**	**.9311261**
polengage_scale	**0**	(omitted)				
male#c.polengage_scale						

1	-.0772258	.7131119	-0.11	0.914	-1.4749	1.320448
race						
1	-.3617132	1.107479	-0.33	0.744	-2.532338	1.808912
2	-.1334373	1.364458	-0.10	0.922	-2.807734	2.540859
3	-.8408615	1.29999	-0.65	0.518	-3.388827	1.707104
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.8775791	1.681607	-0.52	0.602	-4.173479	2.418321
2	-.6678315	2.081802	-0.32	0.748	-4.748107	3.412444
3	.0100971	1.956342	0.01	0.996	-3.824322	3.844517
education	1.239595	1.151714	1.08	0.282	-1.017742	3.496932
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.0364476	1.576735	-0.02	0.982	-3.126811	3.053916
income	.4859998	.9340465	0.52	0.603	-1.344765	2.316764
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.2805289	1.259141	0.22	0.824	-2.187407	2.748464
5.sample_num	.3455837	.1665459	2.08	0.038	.0191596	.6720078
_cons	-.9074492	1.323811	-0.69	0.493	-3.502074	1.687176

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0673166	.3274268	0.21	0.837	-.574428	.7090613

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0293206	.2257001	-0.13	0.897	-.4716846	.4130435

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1241158	.1285666	-0.97	0.334	-.3761017	.1278701

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.2131909	.0783738	-2.72	0.007	-.3668007	-.0595811

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.2937303	.1254976	-2.34	0.019	-.5397011	-.0477595

Multiple-imputation estimates

DF adjustment: **Large sample**

Model F test: **Equal FMI**
Within VCE type: **OIM**

Imputations	=	25
Number of obs	=	968
Average RVI	=	0.0199
Largest FMI	=	0.0197
DF: min	=	61,901.41
avg	=	3455299.17
max	=	4.07e+07
F(18, 1.0e+06)	=	1.75
Prob > F	=	0.0254

dv_wash_hands	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	3.881733	2.2042	1.76	0.078	-.4384196	8.201885
polengage_scale	5.44455	3.027245	1.80	0.072	-.4887975	11.3779
c.auth_scale#c.polengage_scale	-6.174252	3.075304	-2.01	0.045	-12.20174	-.146767
age	.164362	2.330878	0.07	0.944	-4.404079	4.732803
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	.0184544	3.237128	0.01	0.995	-6.326246	6.363154
1.male	.6552263	.933923	0.70	0.483	-1.175233	2.485685
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-2.149658	1.375362	-1.56	0.118	-4.845351	.5460351
race						
1	1.396507	1.394915	1.00	0.317	-1.33748	4.130495
2	.3340311	1.943448	0.17	0.864	-3.475091	4.143153
3	3.68102	2.463874	1.49	0.135	-1.148121	8.510162
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-1.576437	2.3121	-0.68	0.495	-6.108096	2.955222
2	.3663968	3.29235	0.11	0.911	-6.086614	6.819408
3	-3.89701	3.546722	-1.10	0.272	-10.84854	3.05452
education	-3.049291	2.078811	-1.47	0.142	-7.123699	1.025117
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	3.208302	3.114626	1.03	0.303	-2.896372	9.312976
income	1.327537	1.706063	0.78	0.436	-2.016291	4.671365
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.1657744	2.487846	0.07	0.947	-4.71034	5.041889
5.sample_num	-.5625921	.3201524	-1.76	0.079	-1.190083	.0648988
_cons	-.651804	1.881232	-0.35	0.729	-4.338959	3.035351

(1) **- 1bn._at + 6._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.5330358	.3168854	1.68	0.093	-.0880481	1.15412

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.2323118	.1773315	1.31	0.190	-.1152515	.5798751

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.049298	.0550904	0.89	0.371	-.0586772	.1572731

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0303668	.028623	-1.06	0.289	-.0864668	.0257332

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0695745	.0535968	-1.30	0.194	-.1746224	.0354733

Multiple-imputation estimates — Imputations = 25
Number of obs = 966
Average RVI = 0.0146
Largest FMI = 0.0414
DF adjustment: Large sample — DF: min = 14,065.98
avg = 5621724.87
max = 6.24e+07
Model F test: Equal FMI — F(18, 1.8e+06) = 2.10
Within VCE type: OIM — Prob > F = 0.0040

dv_avoid_dine	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	3.416417	1.576424	2.17	0.030	.3266829	6.50615
polengage_scale	2.948861	2.27126	1.30	0.194	-1.502744	7.400467
c.auth_scale#c.polengage_scale	-6.549303	2.124065	-3.08	0.002	-10.71239	-2.386211
age	1.328065	1.490046	0.89	0.373	-1.592372	4.248501
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-2.245605	2.017189	-1.11	0.266	-6.199228	1.708018
1.male	-.7124046	.61397	-1.16	0.246	-1.915764	.4909547
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.5318217	.8506756	0.63	0.532	-1.135474	2.199117

race						
1	-1.79746	1.267831	-1.42	0.156	-4.282373	.6874522
2	-.9593021	1.745571	-0.55	0.583	-4.380642	2.462038
3	-2.360908	1.493427	-1.58	0.114	-5.287992	.5661759
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	2.457803	1.779763	1.38	0.167	-1.030528	5.946133
2	2.257971	2.581347	0.87	0.382	-2.801687	7.317629
3	3.377872	2.157029	1.57	0.117	-.8499249	7.605669
education	-.1895378	1.335873	-0.14	0.887	-2.807819	2.428743
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.4918169	1.889174	0.26	0.795	-3.21097	4.194604
income	.5032574	1.052755	0.48	0.633	-1.560239	2.566754
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-1.275153	1.453907	-0.88	0.380	-4.125003	1.574698
5.sample_num	-.0159836	.2111628	-0.08	0.940	-.4298559	.3978887
_cons	.3503466	1.587791	0.22	0.825	-2.76167	3.462364

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.662546	.2144832	3.09	0.002	.2421666	1.082925

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.3545228	.2021989	1.75	0.080	-.0417798	.7508254

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0251132	.1087226	0.23	0.817	-.1879792	.2382056

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.2224308	.0659689	-3.37	0.001	-.3517275	-.0931342

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.4024763	.1141169	-3.53	0.000	-.6261412	-.1788114

Multiple-imputation estimates

	Imputations	=	25
	Number of obs	=	961
	Average RVI	=	0.0086
	Largest FMI	=	0.0350
DF adjustment: **Large sample**	DF: min	=	19,645.96
	avg	=	6230369.18
	max	=	5.29e+07
Model F test: **Equal FMI**	F(18, 5.3e+06)	=	2.33
Within VCE type: **OIM**	Prob > F	=	0.0011

dv_cancel_event	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	1.155106	1.465769	0.79	0.431	-1.717749	4.027962
polengage_scale	2.567208	2.055824	1.25	0.212	-1.462156	6.596573
c.auth_scale#c.polengage_scale	-2.734159	1.94433	-1.41	0.160	-6.544976	1.076659
age	-.1900027	1.398324	-0.14	0.892	-2.930669	2.550664
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	.2213979	1.880352	0.12	0.906	-3.464025	3.906821
1.male	-.6201605	.5887538	-1.05	0.292	-1.774097	.5337758
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.001309	.8084881	0.00	0.999	-1.583299	1.585917
race						
1	-.1342258	1.123109	-0.12	0.905	-2.335515	2.067064
2	-.951358	1.414823	-0.67	0.501	-3.724396	1.82168
3	.8035768	1.422656	0.56	0.572	-1.984823	3.591977
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.2568032	1.632433	0.16	0.875	-2.94277	3.456376
2	1.616671	2.137486	0.76	0.449	-2.572787	5.80613
3	-1.266628	2.029461	-0.62	0.533	-5.244399	2.711143
education	.7801881	1.336366	0.58	0.559	-1.839046	3.399422
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.1493512	1.858591	-0.08	0.936	-3.792134	3.493431
income	1.953005	1.123203	1.74	0.082	-.2485677	4.154578
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-1.636941	1.503166	-1.09	0.276	-4.583236	1.309354
5.sample_num	-.5145625	.1968001	-2.61	0.009	-.9002841	-.1288409
_cons	-.2940846	1.424939	-0.21	0.836	-3.08693	2.498761

(1) **- 1bn._at + 6._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.2182607	.2737131	0.80	0.425	-.318207	.7547285

(1) **- 2._at + 7._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0807944	.1721691	0.47	0.639	-.2566508	.4182396

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0322482	.0885431	-0.36	0.716	-.2057894	.141293

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1200763	.0536674	-2.24	0.025	-.2252625	-.0148901

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1876489	.0894126	-2.10	0.036	-.3628943	-.0124035

Multiple-imputation estimates — Imputations = 25
Number of obs = 965
Average RVI = 0.0008
Largest FMI = 0.0015
DF adjustment: Large sample — DF: min = 1.05e+07
avg = 1.60e+08
max = 6.79e+08
Model F test: Equal FMI — F(18, 5.6e+08) = 1.77
Within VCE type: OIM — Prob > F = 0.0227

dv_sanitize	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	3.677641	2.167623	1.70	0.090	-.570823	7.926105
polengage_scale	.1005645	4.212497	0.02	0.981	-8.155777	8.356906
c.auth_scale#c.polengage_scale	-6.237128	2.978072	-2.09	0.036	-12.07404	-.400214
age	2.142184	2.41289	0.89	0.375	-2.586994	6.871362
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-2.27122	3.229781	-0.70	0.482	-8.601475	4.059036
1.male	-.4350974	.8947384	-0.49	0.627	-2.188752	1.318558
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.5629922	1.272786	-0.44	0.658	-3.057607	1.931623
race						
1	-2.475819	2.660989	-0.93	0.352	-7.691261	2.739623
2	-4.928679	2.876291	-1.71	0.087	-10.56611	.7087474
3	-4.270076	2.961637	-1.44	0.149	-10.07478	1.534627
polengage_scale	0	(omitted)				
race#c.polengage_scale						

1	2.959349	3.572164	0.83	0.407	-4.041964	9.960661
2	6.568452	4.076531	1.61	0.107	-1.421403	14.55831
3	7.051939	4.376721	1.61	0.107	-1.526277	15.63016
education	-1.193674	1.963811	-0.61	0.543	-5.042672	2.655324
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	2.115755	2.84521	0.74	0.457	-3.460753	7.692263
income	-.4851826	1.591896	-0.30	0.761	-3.605241	2.634876
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	1.734598	2.263417	0.77	0.443	-2.701618	6.170814
5.sample_num	-.5054533	.3134446	-1.61	0.107	-1.119794	.1088869
_cons	3.184361	3.051602	1.04	0.297	-2.796669	9.165392

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.3669956	.2954702	1.24	0.214	-.2121153	.9461064

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.1626196	.1400843	1.16	0.246	-.1119406	.4371797

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.033281	.0511083	0.65	0.515	-.0668894	.1334514

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0513304	.0350101	-1.47	0.143	-.1199489	.0172882

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1291032	.0828463	-1.56	0.119	-.291479	.0332726

```
27 .
28 . local vars dv_distancing dv_masking

29 . foreach var of local vars {
  2. mi estimate: regress `var' c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.rac
> num
  3. quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1)) coeflegend post
  4.     lincom _b[6._at]-_b[1._at]
  5.     lincom _b[7._at]-_b[2._at]
  6.     lincom _b[8._at]-_b[3._at]
  7.     lincom _b[9._at]-_b[4._at]
  8.     lincom _b[10._at]-_b[5._at]
  9. }

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        964
                                                Average RVI       =     0.0118
                                                Largest FMI       =     0.0170
                                                Complete DF       =        945
DF adjustment:   Small sample                   DF:     min       =     916.98
                                                        avg       =     933.96
                                                        max       =     942.49
Model F test:       Equal FMI                   F(  18,  942.6)   =       2.44
Within VCE type:          OLS                   Prob > F          =     0.0007
```

dv_distancing	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.1371635	.2477203	0.55	0.580	-.3489837	.6233107
polengage_scale	.3444893	.3065793	1.12	0.261	-.2571747	.9461532
c.latentnsc2#c.polengage_scale	-.1533625	.3177083	-0.48	0.629	-.7768601	.4701351
age	-.2094411	.2105413	-0.99	0.320	-.6226255	.2037432
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	.1040798	.2749476	0.38	0.705	-.4355021	.6436618
1.male	-.0721333	.0896984	-0.80	0.421	-.2481651	.1038985
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0299257	.1187276	-0.25	0.801	-.2629275	.203076
race						
1	.2241386	.1688654	1.33	0.185	-.1072593	.5555365
2	.1390547	.2147096	0.65	0.517	-.2823151	.5604244
3	.1260821	.2064838	0.61	0.542	-.2791454	.5313097
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.4543349	.2393619	-1.90	0.058	-.9240872	.0154173
2	-.2360823	.3117005	-0.76	0.449	-.8478031	.3756386
3	-.2222951	.2931732	-0.76	0.449	-.7976618	.3530716
education	-.0761176	.19268	-0.40	0.693	-.4542513	.3020161
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.1230533	.2596666	0.47	0.636	-.3865432	.6326498
income	.064817	.152365	0.43	0.671	-.2342076	.3638415
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.0888448	.2007836	0.44	0.658	-.3051991	.4828887
5.sample_num	-.0260713	.0276793	-0.94	0.346	-.0803925	.0282499

_cons	.7176473	.2190431	3.28	0.001	.2877764	1.147518

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0997179	.1800927	0.55	0.580	-.2537099	.4531457

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0718442	.1248287	0.58	0.565	-.1731293	.3168177

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0439706	.0734823	0.60	0.550	-.1002368	.188178

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.016097	.043425	0.37	0.711	-.0691236	.1013175

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0117767	.0709959	-0.17	0.868	-.1511046	.1275512

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 964
Average RVI = 0.0053
Largest FMI = 0.0142
Complete DF = 945
DF adjustment: Small sample — DF: min = 922.49
avg = 939.11
max = 942.69
Model F test: Equal FMI — F(18, 942.9) = 1.42
Within VCE type: OLS — Prob > F = 0.1139

dv_masking	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.1312255	.134764	0.97	0.330	-.1332466	.3956976
polengage_scale	.3025868	.168778	1.79	0.073	-.0286401	.6338136
c.latentnsc2#c.polengage_scale	-.2712315	.1730562	-1.57	0.117	-.6108516	.0683886
age	.0945075	.1145883	0.82	0.410	-.1303704	.3193853
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.1423844	.1497177	-0.95	0.342	-.4362034	.1514346

1.male	.0029576	.0490747	0.06	0.952	-.0933506	.0992659
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0125762	.0649032	-0.19	0.846	-.1399477	.1147952
race						
1	.0057851	.094653	0.06	0.951	-.1799706	.1915408
2	-.1691183	.1197863	-1.41	0.158	-.404198	.0659615
3	-.1208686	.1139777	-1.06	0.289	-.3445499	.1028127
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.0448525	.1334839	-0.34	0.737	-.306817	.2171121
2	.2191268	.17279	1.27	0.205	-.1199743	.5582279
3	.14018	.1617242	0.87	0.386	-.1772102	.4575701
education	.1352649	.1042614	1.30	0.195	-.0693466	.3398764
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.1928565	.1407595	-1.37	0.171	-.4690949	.083382
income	-.0350485	.0836867	-0.42	0.675	-.1992827	.1291857
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.110008	.1102682	1.00	0.319	-.1063919	.3264079
5.sample_num	-.034625	.0150766	-2.30	0.022	-.0642127	-.0050373
_cons	.8074749	.1207388	6.69	0.000	.5705264	1.044423

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.095401	.0979734	0.97	0.330	-.0968697	.2876716

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0461046	.0678774	0.68	0.497	-.0871032	.1793125

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0031917	.0399308	-0.08	0.936	-.0815551	.0751717

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.052488	.0236681	-2.22	0.027	-.0989362	-.0060398

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**-.1017844**	**.0387875**	**-2.62**	**0.009**	**-.1779039**	**-.0256648**

```
30 .
31 . local vars dv_travel dv_work_home dv_wash_hands dv_avoid_dine dv_cancel_event dv_sanitize

32 . foreach var of local vars {
  2. mi estimate, cmdok: firthlogit `var' c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_s
>  i.sample_num
  3. quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1) race=1 sample_num=2) atmeans post e
  4.     lincom _b[6._at]-_b[1._at]
  5.     lincom _b[7._at]-_b[2._at]
  6.     lincom _b[8._at]-_b[3._at]
  7.     lincom _b[9._at]-_b[4._at]
  8.     lincom _b[10._at]-_b[5._at]
  9. }
```

Multiple-imputation estimates — Imputations = **25**; Number of obs = **958**; Average RVI = **0.0069**; Largest FMI = **0.0266**

DF adjustment: **Large sample** — DF: min = **34,052.87**; avg = **1.20e+07**; max = **1.06e+08**

Model F test: **Equal FMI** — F(**18, 8.2e+06**) = **3.14**

Within VCE type: **OIM** — Prob > F = **0.0000**

dv_travel	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	**-3.232433**	**1.599559**	**-2.02**	**0.043**	**-6.367512**	**-.0973546**
polengage_scale	**-1.530202**	**2.122266**	**-0.72**	**0.471**	**-5.689785**	**2.629382**
c.latentnsc2#c.polengage_scale	**3.589252**	**2.077623**	**1.73**	**0.084**	**-.4828155**	**7.661319**
age	**1.019223**	**1.300955**	**0.78**	**0.433**	**-1.530603**	**3.569049**
polengage_scale	**0**	(omitted)				
c.age#c.polengage_scale	**-2.125105**	**1.735791**	**-1.22**	**0.221**	**-5.527193**	**1.276983**
1.male	**-.4615533**	**.5609672**	**-0.82**	**0.411**	**-1.561029**	**.6379225**
polengage_scale	**0**	(omitted)				
male#c.polengage_scale						
1	**.063778**	**.7578453**	**0.08**	**0.933**	**-1.421572**	**1.549128**
race						
1	**-1.391416**	**1.200547**	**-1.16**	**0.246**	**-3.744472**	**.9616407**
2	**-.3947288**	**1.459018**	**-0.27**	**0.787**	**-3.25437**	**2.464913**
3	**-1.533962**	**1.399606**	**-1.10**	**0.273**	**-4.277182**	**1.209258**
polengage_scale	**0**	(omitted)				
race#c.polengage_scale						
1	**1.549614**	**1.694096**	**0.91**	**0.360**	**-1.770782**	**4.870011**
2	**.5280102**	**2.119712**	**0.25**	**0.803**	**-3.626572**	**4.682593**
3	**1.891024**	**2.00106**	**0.95**	**0.345**	**-2.031022**	**5.813069**
education	**.0756339**	**1.237102**	**0.06**	**0.951**	**-2.349096**	**2.500364**
polengage_scale	**0**	(omitted)				
c.education#c.polengage_scale	**1.096901**	**1.702001**	**0.64**	**0.519**	**-2.239015**	**4.432817**
income	**2.256605**	**1.067409**	**2.11**	**0.035**	**.1644475**	**4.348761**

polengage_scale	**0**	(omitted)				
c.income#c.polengage_scale	**-1.318148**	**1.419196**	**-0.93**	**0.353**	**-4.099813**	**1.463517**
5.sample_num	**-.3323408**	**.1811104**	**-1.84**	**0.067**	**-.6873107**	**.0226292**
_cons	**2.280855**	**1.512434**	**1.51**	**0.132**	**-.683477**	**5.245186**

(1) **- 1bn._at + 6._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.4835603**	**.1951844**	**-2.48**	**0.013**	**-.8661147**	**-.1010059**

(1) **- 2._at + 7._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.337932**	**.1430005**	**-2.36**	**0.018**	**-.6182077**	**-.0576562**

(1) **- 3._at + 8._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1930924	.0821404	-2.35	0.019	-.3540847	-.0321001

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0651481	.0484121	-1.35	0.178	-.160034	.0297378

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0378495	.0709358	0.53	0.594	-.1011821	.1768811

Multiple-imputation estimates

DF adjustment: **Large sample**

Model F test: **Equal FMI**
Within VCE type: **OIM**

Imputations	=	25
Number of obs	=	951
Average RVI	=	0.0108
Largest FMI	=	0.0269
DF: min	=	33,400.63
avg	=	1.42e+07
max	=	1.20e+08
F(18, 3.4e+06)	=	4.66
Prob > F	=	0.0000

dv_work_home	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.145343	1.493047	0.10	0.922	-2.780974	3.07166
polengage_scale	2.906799	1.994956	1.46	0.145	-1.003248	6.816846
c.latentnsc2#c.polengage_scale	-1.591135	1.945486	-0.82	0.413	-5.404218	2.221948
age	.4915771	1.277133	0.38	0.700	-2.011558	2.994713
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-1.71375	1.682126	-1.02	0.308	-5.010656	1.583157
1.male	-.0992429	.5305311	-0.19	0.852	-1.139065	.9405791
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0920992	.7131889	-0.13	0.897	-1.489924	1.305726
race						
1	-.3343698	1.108494	-0.30	0.763	-2.506982	1.838243
2	-.1134165	1.368791	-0.08	0.934	-2.796205	2.569372
3	-.8411803	1.300788	-0.65	0.518	-3.39071	1.70835
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.9370659	1.68477	-0.56	0.578	-4.239164	2.365033
2	-.7075431	2.08754	-0.34	0.735	-4.799065	3.383979
3	-.0080363	1.958373	-0.00	0.997	-3.846438	3.830366
education	1.282263	1.151212	1.11	0.265	-.9740884	3.538615
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.1109933	1.577451	-0.07	0.944	-3.202758	2.980772

income	.4773858	.934419	0.51	0.609	-1.354108	2.30888
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.3028606	1.260005	0.24	0.810	-2.166767	2.772488
5.sample_num	.3462635	.1667237	2.08	0.038	.0194909	.673036
_cons	-.8948933	1.341794	-0.67	0.505	-3.524764	1.734977

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0263936	.2709371	0.10	0.922	-.5046334	.5574205

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0456951	.1859652	-0.25	0.806	-.4101803	.31879

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1154648	.1056612	-1.09	0.274	-.3225569	.0916272

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1798844	.0633424	-2.84	0.005	-.3040332	-.0557357

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.2367848	.0994666	-2.38	0.017	-.4317359	-.0418338

Multiple-imputation estimates — Imputations = 25
Number of obs = 968
Average RVI = 0.0199
Largest FMI = 0.0204
DF adjustment: Large sample — DF: min = 57,583.14
avg = 2332518.45
max = 2.69e+07
Model F test: Equal FMI — F(18, 1.0e+06) = 1.73
Within VCE type: OIM — Prob > F = 0.0284

dv_wash_hands	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	4.212677	2.508278	1.68	0.093	-.7034587	9.128812
polengage_scale	5.563702	3.046507	1.83	0.068	-.4073982	11.5348
c.latentnsc2#c.polengage_scale	-6.730723	3.498951	-1.92	0.054	-13.58854	.1270967
age	-.0145662	2.360327	-0.01	0.995	-4.640725	4.611593
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	.2968236	3.273959	0.09	0.928	-6.120065	6.713712
1.male	.6893527	.9376776	0.74	0.462	-1.148465	2.527171
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-2.206994	1.380894	-1.60	0.110	-4.91353	.4995411
race						
1	1.3856	1.388083	1.00	0.318	-1.334996	4.106196
2	.2132145	1.968593	0.11	0.914	-3.645196	4.071625
3	3.624544	2.451462	1.48	0.139	-1.180268	8.429357
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-1.57116	2.293733	-0.68	0.493	-6.066819	2.924498
2	.5437936	3.330017	0.16	0.870	-5.983056	7.070643
3	-3.817847	3.522617	-1.08	0.278	-10.72213	3.086435
education	-3.100307	2.071649	-1.50	0.135	-7.160679	.9600644
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	3.257664	3.107571	1.05	0.295	-2.833184	9.348511
income	1.2755	1.719434	0.74	0.458	-2.094535	4.645535
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.2553397	2.501856	0.10	0.919	-4.648237	5.158917
5.sample_num	-.56942	.3207463	-1.78	0.076	-1.198075	.0592352
_cons	-.6962303	1.88747	-0.37	0.712	-4.395612	3.003151

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.4894208	.3261486	1.50	0.133	-.1498186	1.12866

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.2053451	.1741356	1.18	0.238	-.1359544	.5466446

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0402664	.0503473	0.80	0.424	-.0584125	.1389452

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0229785	.0207299	-1.11	0.268	-.0636083	.0176513

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0431626	.0289424	-1.49	0.136	-.0998887	.0135635

Multiple-imputation estimates

DF adjustment: **Large sample**

Model F test: **Equal FMI**
Within VCE type: **OIM**

Imputations	=		25
Number of obs	=		966
Average RVI	=		0.0142
Largest FMI	=		0.0414
DF: min	=		14,073.66
avg	=		7267992.83
max	=		8.48e+07
F(18, 1.9e+06)	=		2.19
Prob > F	=		0.0026

dv_avoid_dine	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	3.028673	1.797097	1.69	0.092	-.493572	6.550917
polengage_scale	3.048437	2.307622	1.32	0.186	-1.474435	7.571308
c.latentnsc2#c.polengage_scale	-6.582338	2.406403	-2.74	0.006	-11.2988	-1.865875
age	1.37885	1.504129	0.92	0.359	-1.569189	4.326889
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-2.18563	2.0345	-1.07	0.283	-6.17318	1.80192
1.male	-.6588015	.6143846	-1.07	0.284	-1.862973	.5453704
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.4580341	.8522357	0.54	0.591	-1.212319	2.128387
race						
1	-1.75666	1.264843	-1.39	0.165	-4.235716	.7223958
2	-.9295795	1.746669	-0.53	0.595	-4.353069	2.49391
3	-2.341657	1.489348	-1.57	0.116	-5.260746	.5774306
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	2.372896	1.776839	1.34	0.182	-1.109698	5.85549
2	2.219545	2.583982	0.86	0.390	-2.845272	7.284361
3	3.322816	2.151207	1.54	0.122	-.8935636	7.539195
education	-.1789235	1.331117	-0.13	0.893	-2.787882	2.430035
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.4156103	1.888164	0.22	0.826	-3.285198	4.116418

income	.4569244	1.054684	0.43	0.665	-1.610352	2.524201
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-1.178271	1.45767	-0.81	0.419	-4.035498	1.678957
5.sample_num	-.016999	.2119152	-0.08	0.936	-.4323459	.3983479
_cons	.4863037	1.609374	0.30	0.763	-2.668014	3.640622

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.4891927	.2538382	1.93	0.054	-.0083211	.9867064

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.2129505	.1933688	1.10	0.271	-.1660453	.5919463

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0333081	.0885223	-0.38	0.707	-.2068087	.1401925

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.1870068	.0469403	-3.98	0.000	-.2790081	-.0950056

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.2729349	.0746455	-3.66	0.000	-.4192374	-.1266323

Multiple-imputation estimates Imputations = 25
Number of obs = 961
Average RVI = 0.0085
Largest FMI = 0.0352
DF adjustment: Large sample DF: min = 19,518.58
avg = 4789590.97
max = 3.78e+07
Model F test: Equal FMI F(18, 5.3e+06) = 2.39
Within VCE type: OIM Prob > F = 0.0008

dv_cancel_event	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.8948201	1.664026	0.54	0.591	-2.36661	4.15625
polengage_scale	2.624412	2.085457	1.26	0.208	-1.463033	6.711857
c.latentnsc2#c.polengage_scale	-2.741818	2.193488	-1.25	0.211	-7.040976	1.55734
age	-.1496076	1.410341	-0.11	0.916	-2.913826	2.61461
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	.2493921	1.895329	0.13	0.895	-3.465387	3.964171
1.male	-.5978858	.5892521	-1.01	0.310	-1.752799	.5570275
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0320773	.8097156	-0.04	0.968	-1.619091	1.554936
race						
1	-.109337	1.123593	-0.10	0.922	-2.311575	2.092901
2	-.9283852	1.420595	-0.65	0.513	-3.712735	1.855965
3	.7901423	1.421303	0.56	0.578	-1.995606	3.575891
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.2069982	1.633437	0.13	0.899	-2.994543	3.408539
2	1.592649	2.144745	0.74	0.458	-2.611037	5.796336
3	-1.257541	2.028013	-0.62	0.535	-5.232473	2.717391
education	.8106999	1.334892	0.61	0.544	-1.805646	3.427045
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.221774	1.859381	-0.12	0.905	-3.866104	3.422556
income	1.943116	1.124675	1.73	0.084	-.2613431	4.147576
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-1.599239	1.505798	-1.06	0.288	-4.550696	1.352218
5.sample_num	-.5192438	.197249	-2.63	0.008	-.9058451	-.1326424
_cons	-.2069619	1.444823	-0.14	0.886	-3.038778	2.624855

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.1266186	.2428811	0.52	0.602	-.3494196	.6026567

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0261041	.1440801	0.18	0.856	-.2562876	.3084958

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.0511398**	**.0699127**	**-0.73**	**0.464**	**-.1881661**	**.0858866**

(1) **- 4._at + 9._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.1060282**	**.0395622**	**-2.68**	**0.007**	**-.1835687**	**-.0284876**

(1) **- 5._at + 10._at = 0**

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	**-.1428799**	**.0603273**	**-2.37**	**0.018**	**-.2611192**	**-.0246405**

Multiple-imputation estimates — Imputations = **25**; Number of obs = **965**; Average RVI = **0.0008**; Largest FMI = **0.0014**

DF adjustment: **Large sample** — DF: min = **1.19e+07**; avg = **1.72e+08**; max = **6.44e+08**

Model F test: **Equal FMI** — F(**18, 6.0e+08**) = **1.82**

Within VCE type: **OIM** — Prob > F = **0.0181**

dv_sanitize	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	**3.562867**	**2.502009**	**1.42**	**0.154**	**-1.34098**	**8.466714**
polengage_scale	**.5789648**	**4.262939**	**0.14**	**0.892**	**-7.776243**	**8.934172**
c.latentnsc2#c.polengage_scale	**-6.781335**	**3.423487**	**-1.98**	**0.048**	**-13.49125**	**-.0714226**
age	**2.096238**	**2.438788**	**0.86**	**0.390**	**-2.683698**	**6.876174**
polengage_scale	**0**	(omitted)				
c.age#c.polengage_scale	**-2.042827**	**3.266578**	**-0.63**	**0.532**	**-8.445202**	**4.359547**
1.male	**-.3386625**	**.8933884**	**-0.38**	**0.705**	**-2.089672**	**1.412347**
polengage_scale	**0**	(omitted)				
male#c.polengage_scale						
1	**-.6960273**	**1.273303**	**-0.55**	**0.585**	**-3.191656**	**1.799601**
race						
1	**-2.373739**	**2.626221**	**-0.90**	**0.366**	**-7.521038**	**2.773559**
2	**-4.866086**	**2.845819**	**-1.71**	**0.087**	**-10.44379**	**.7116166**
3	**-4.189604**	**2.927185**	**-1.43**	**0.152**	**-9.926781**	**1.547573**
polengage_scale	**0**	(omitted)				
race#c.polengage_scale						
1	**2.789798**	**3.531606**	**0.79**	**0.430**	**-4.132022**	**9.711617**
2	**6.514106**	**4.045423**	**1.61**	**0.107**	**-1.414778**	**14.44299**
3	**6.918981**	**4.332179**	**1.60**	**0.110**	**-1.571935**	**15.4099**
education	**-1.132798**	**1.956618**	**-0.58**	**0.563**	**-4.9677**	**2.702103**
polengage_scale	**0**	(omitted)				
c.education#c.polengage_scale	**1.939743**	**2.848303**	**0.68**	**0.496**	**-3.642828**	**7.522314**

income	-.4951722	1.602892	-0.31	0.757	-3.636784	2.646439
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	1.811262	2.280318	0.79	0.427	-2.65808	6.280604
5.sample_num	-.5145962	.3144326	-1.64	0.102	-1.130873	.1016803
_cons	3.09006	3.0686	1.01	0.314	-2.924285	9.104405

(1) - 1bn._at + 6._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.2890783	.2889296	1.00	0.317	-.2772133	.8553699

(1) - 2._at + 7._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.111228	.1262542	0.88	0.378	-.1362258	.3586817

(1) - 3._at + 8._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	.0074646	.0427463	0.17	0.861	-.0763167	.0912459

(1) - 4._at + 9._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.0510979	.0250822	-2.04	0.042	-.1002581	-.0019376

(1) - 5._at + 10._at = 0

	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
(1)	-.091144	.0472829	-1.93	0.054	-.1838167	.0015287

```
19 .
20 .
21 .
end of do-file
```

```
22 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000002.tmp"

23 .
24 .
25 .
26 .
end of do-file

27 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000003.tmp"

28 .
29 .
30 .
31 . ***** SM2 - Latent NSC with/without Covariances Modeled *****
32 .
33 . * Reverse-Coded Covariances Modeled
34 . mi estimate: regress z_dv_concern c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
>  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        965
                                                Average RVI       =     0.0112
                                                Largest FMI       =     0.0415
                                                Complete DF       =        946
DF adjustment:   Small sample                   DF:     min       =     850.29
                                                        avg       =     926.16
                                                        max       =     943.25
Model F test:       Equal FMI                   F(  18,  943.7)   =       2.70
Within VCE type:          OLS                   Prob > F          =     0.0002
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.045829	.0370731	-1.24	0.217	-.1185843	.0269262
z_polengage_scale	.0029868	.1936837	0.02	0.988	-.3771217	.3830952
c.z_latentnsc2#c.z_polengage_scale	-.0923728	.0479984	-1.92	0.055	-.1865689	.0018232
age	.045498	.1692745	0.27	0.788	-.2867003	.3776963
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1890517	.2052814	-0.92	0.357	-.5919134	.21381
1.male	-.1672555	.070208	-2.38	0.017	-.3050376	-.0294734
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.120961	.0869399	1.39	0.164	-.0496573	.2915793
race						
1	-.1674628	.1474239	-1.14	0.256	-.4567855	.1218599
2	.2456624	.1850288	1.33	0.185	-.117476	.6088007
3	.0587673	.1791187	0.33	0.743	-.2927629	.4102975
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0884646	.17277	0.51	0.609	-.2506052	.4275345
2	-.1509971	.2296966	-0.66	0.511	-.6018358	.2998417
3	-.1272016	.2079613	-0.61	0.541	-.5353463	.2809431
education	.3574676	.155128	2.30	0.021	.0530287	.6619065
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.1408716	.186989	0.75	0.451	-.2260966	.5078398
income	-.2664761	.131534	-2.03	0.043	-.5246143	-.008338

z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0345328	.1466231	-0.24	0.814	-.3222848	.2532193
5.sample_num	-.0620829	.0804159	-0.77	0.440	-.2198981	.0957323
_cons	-.1848749	.1799615	-1.03	0.305	-.5380511	.1683013

35 . mi estimate: regress z_dv_behavior c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
> e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates	Imputations	=	25
Linear regression	Number of obs	=	970
	Average RVI	=	0.0100
	Largest FMI	=	0.0224
	Complete DF	=	951
DF adjustment: Small sample	DF: min	=	910.40
	avg	=	933.13
	max	=	948.29
Model F test: Equal FMI	F(18, 948.7)	=	4.11
Within VCE type: OLS	Prob > F	=	0.0000

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.0792841	.0294006	-2.70	0.007	-.1369819	-.0215864
z_polengage_scale	.1522316	.1624136	0.94	0.349	-.1665041	.4709673
c.z_latentnsc2#c.z_polengage_scale	-.0476964	.0382142	-1.25	0.212	-.1226906	.0272977
age	-.124416	.1343222	-0.93	0.355	-.3880218	.1391899
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1727281	.1631452	-1.06	0.290	-.4928966	.1474404
1.male	-.2315128	.0565879	-4.09	0.000	-.3425658	-.1204597
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.0172722	.0704937	0.25	0.806	-.1210699	.1556143
race						
1	-.2025207	.1174573	-1.72	0.085	-.4330358	.0279944
2	-.0287809	.1472701	-0.20	0.845	-.3178022	.2602403
3	-.1124678	.14412	-0.78	0.435	-.3953133	.1703776
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	-.0058489	.1423268	-0.04	0.967	-.2851647	.2734668
2	.1137387	.18537	0.61	0.540	-.2500516	.4775289
3	.0824457	.1747175	0.47	0.637	-.2604477	.4253391
education	.2247866	.1271918	1.77	0.078	-.0248285	.4744017
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0546148	.1534042	0.36	0.722	-.2464409	.3556705
income	.3569805	.1071197	3.33	0.001	.1467532	.5672078
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0456332	.1196956	-0.38	0.703	-.2805446	.1892783
5.sample_num	-.0954953	.0647909	-1.47	0.141	-.2226464	.0316557
_cons	.3608952	.1436351	2.51	0.012	.0790087	.6427817

```
36 .
37 . * Reverse-Coded Covariances Not Modeled
38 . mi estimate: regress z_dv_concern c.z_latentnsc1##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
   >  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =        965
                                              Average RVI       =     0.0110
                                              Largest FMI       =     0.0412
                                              Complete DF       =        946
DF adjustment:   Small sample                 DF:     min       =     851.26
                                                      avg       =     926.17
                                                      max       =     943.27
Model F test:       Equal FMI                 F(  18,  943.7)   =       2.73
Within VCE type:          OLS                 Prob > F          =     0.0001
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc1	-.0429911	.0372966	-1.15	0.249	-.116185	.0302027
z_polengage_scale	-.00401	.1938414	-0.02	0.983	-.384428	.3764081
c.z_latentnsc1#c.z_polengage_scale	-.097924	.0474124	-2.07	0.039	-.19097	-.004878
age	.0464319	.1698857	0.27	0.785	-.2869659	.3798297
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.177821	.2055999	-0.86	0.387	-.5813076	.2256657
1.male	-.1666614	.0701625	-2.38	0.018	-.3043542	-.0289687
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.1201484	.0869243	1.38	0.167	-.0504393	.2907362
race						
1	-.1731218	.1474187	-1.17	0.241	-.4624344	.1161907
2	.242206	.1849397	1.31	0.191	-.1207576	.6051696
3	.0581334	.1790562	0.32	0.746	-.2932742	.409541
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0868505	.1728133	0.50	0.615	-.2523045	.4260055
2	-.1445204	.2298491	-0.63	0.530	-.5956577	.306617
3	-.1216227	.2079595	-0.58	0.559	-.5297635	.2865182
education	.3549771	.1550208	2.29	0.022	.0507486	.6592056
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.1436836	.1869367	0.77	0.442	-.2231819	.5105492
income	-.2637514	.1315332	-2.01	0.045	-.5218879	-.0056148
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0305286	.1465331	-0.21	0.835	-.3181042	.2570471
5.sample_num	-.0634102	.080407	-0.79	0.431	-.2212079	.0943875
_cons	-.1811377	.1800058	-1.01	0.315	-.534401	.1721255

```
39 . mi estimate: regress z_dv_behavior c.z_latentnsc1##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
   > e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        970
                                                Average RVI       =     0.0098
                                                Largest FMI       =     0.0217
                                                Complete DF       =        951
DF adjustment:   Small sample                   DF:     min       =     912.05
                                                        avg       =     933.62
                                                        max       =     948.18
Model F test:       Equal FMI                   F(  18,  948.7)   =       4.30
Within VCE type:          OLS                   Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc1	-.1003899	.0296605	-3.38	0.001	-.1585976	-.0421821
z_polengage_scale	.1608312	.1624068	0.99	0.322	-.157891	.4795535
c.z_latentnsc1#c.z_polengage_scale	-.0284991	.0380533	-0.75	0.454	-.1031775	.0461794
age	-.0925301	.1347216	-0.69	0.492	-.3569197	.1718596
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1839738	.1634262	-1.13	0.261	-.5046936	.1367461
1.male	-.228436	.0564658	-4.05	0.000	-.3392493	-.1176227
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.0163823	.0703635	0.23	0.816	-.1217042	.1544689
race						
1	-.2090806	.1173156	-1.78	0.075	-.4393174	.0211562
2	-.0248522	.1469463	-0.17	0.866	-.3132376	.2635332
3	-.1144227	.1438321	-0.80	0.427	-.3967024	.167857
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	-.0154252	.1421849	-0.11	0.914	-.2944624	.263612
2	.1010083	.1854209	0.54	0.586	-.2628818	.4648984
3	.0808943	.1745699	0.46	0.643	-.261709	.4234976
education	.2200059	.1269251	1.73	0.083	-.0290855	.4690974
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0475304	.1530954	0.31	0.756	-.2529192	.34798
income	.3617952	.1069395	3.38	0.001	.1519221	.5716683
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0415718	.1194525	-0.35	0.728	-.2760055	.1928619
5.sample_num	-.0959075	.0646799	-1.48	0.138	-.2228407	.0310256
_cons	.3564955	.1434318	2.49	0.013	.075008	.637983

```
40 .
41 .
   end of do-file

42 .
```

```
25 .
26 .
   end of do-file

27 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000004.tmp"

28 .
29 . ***** SM3 - Conditional Sorting Models *****
30 . alpha pid ideo, gen(leftright)

   Test scale = mean(unstandardized items)

   Average interitem covariance:       .06932
   Number of items in the scale:            2
   Scale reliability coefficient:      0.7962

31 .
32 . mi estimate: regress pid c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.race##c.
   > engage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

   Multiple-imputation estimates                   Imputations       =         25
   Linear regression                               Number of obs     =      8,498
                                                   Average RVI       =     0.0034
                                                   Largest FMI       =     0.0517
                                                   Complete DF       =       8424
   DF adjustment:   Small sample                   DF:     min       =   4,245.17
                                                           avg       =   6,466.35
                                                           max       =   8,419.03
   Model F test:       Equal FMI                   F(  73, 8421.5)   =      20.86
   Within VCE type:          OLS                   Prob > F          =     0.0000
```

pid	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.0897892	.0388145	-2.31	0.021	-.1658752	-.0137032
polengage_scale	-.3537108	.0771279	-4.59	0.000	-.5049006	-.2025211
c.auth_scale#c.polengage_scale	.463864	.049705	9.33	0.000	.36643	.5612979
age	-.0107155	.056002	-0.19	0.848	-.1204935	.0990626
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	.1267246	.076316	1.66	0.097	-.0228739	.2763232
1.male	-.0019628	.024615	-0.08	0.936	-.0502144	.0462889
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.0626906	.0320055	1.96	0.050	-.0000481	.1254294
race						
1	.0837544	.0453342	1.85	0.065	-.0051118	.1726206
2	-.2360336	.0529981	-4.45	0.000	-.3399228	-.1321443
3	-.1060578	.0535728	-1.98	0.048	-.2110737	-.0010419
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.0738581	.0615541	-1.20	0.230	-.1945192	.0468031
2	-.0865427	.0746566	-1.16	0.246	-.2328881	.0598026
3	.0152681	.0754684	0.20	0.840	-.1326687	.1632048
education	-.0626397	.046394	-1.35	0.177	-.1535837	.0283044
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.0213024	.060188	-0.35	0.723	-.1392865	.0966817

income	.0108658	.0463675	0.23	0.815	-.0800323	.101764
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.0717825	.0615142	1.17	0.243	-.0488175	.1923825
sample_num						
2	-.0050516	.0221853	-0.23	0.820	-.0485407	.0384375
3	.0291801	.0173281	1.68	0.092	-.0047872	.0631475
4	-.0437849	.0156273	-2.80	0.005	-.0744183	-.0131514
5	.0220598	.0205932	1.07	0.284	-.018308	.0624275
6	.0045933	.0150322	0.31	0.760	-.0248737	.0340603
state_1	-.230669	.1616921	-1.43	0.154	-.5476406	.0863026
state_2	.0065708	.1485051	0.04	0.965	-.2845554	.297697
state_3	-.0168193	.1494389	-0.11	0.910	-.3097755	.2761369
state_4	-.0145429	.1465442	-0.10	0.921	-.3018262	.2727404
state_5	-.1257682	.1451645	-0.87	0.386	-.4103472	.1588108
state_6	-.0441966	.1474948	-0.30	0.764	-.3333421	.2449488
state_7	-.1271575	.1505461	-0.84	0.398	-.4222843	.1679693
state_8	-.1390745	.1598005	-0.87	0.384	-.4523389	.17419
state_9	-.1556977	.1582612	-0.98	0.325	-.465945	.1545495
state_10	-.0581787	.1452719	-0.40	0.689	-.3429682	.2266108
state_11	-.0281813	.1460692	-0.19	0.847	-.3145336	.2581709
state_12	-.0836325	.1563463	-0.53	0.593	-.3901268	.2228617
state_13	-.0841824	.1490564	-0.56	0.572	-.376389	.2080242
state_14	-.0173417	.152659	-0.11	0.910	-.3166086	.2819251
state_15	-.0713394	.1460078	-0.49	0.625	-.3575711	.2148924
state_16	-.021747	.1472066	-0.15	0.883	-.310328	.2668341
state_17	-.0242932	.1495294	-0.16	0.871	-.317427	.2688405
state_18	-.0869049	.1486343	-0.58	0.559	-.3782845	.2044746
state_19	-.0039732	.1494208	-0.03	0.979	-.296894	.2889476
state_20	-.1578873	.1473585	-1.07	0.284	-.4467664	.1309917
state_21	-.1311105	.1473529	-0.89	0.374	-.4199787	.1577577
state_22	-.0843468	.1541815	-0.55	0.584	-.386598	.2179045
state_23	-.0910232	.1461622	-0.62	0.533	-.3775576	.1955111
state_24	-.0269263	.1482098	-0.18	0.856	-.3174736	.2636211
state_25	-.0384024	.1471182	-0.26	0.794	-.3268107	.2500058
state_26	-.011146	.1517087	-0.07	0.941	-.3085508	.2862589
state_27	-.0698182	.1570732	-0.44	0.657	-.377737	.2381007
state_28	-.0497235	.1463794	-0.34	0.734	-.3366836	.2372367
state_29	-.1771732	.1717688	-1.03	0.302	-.5138952	.1595488
state_30	.0322809	.1533888	0.21	0.833	-.2684167	.3329786
state_31	-.09827	.1548936	-0.63	0.526	-.4019173	.2053773
state_32	-.0782717	.1465032	-0.53	0.593	-.3654746	.2089313
state_33	.0858459	.1534672	0.56	0.576	-.2150051	.3866968
state_34	-.0722576	.1494235	-0.48	0.629	-.3651836	.2206684
state_35	-.1423965	.1453817	-0.98	0.327	-.4274015	.1426084
state_36	-.0806327	.1460257	-0.55	0.581	-.3668995	.2056341
state_37	-.0151729	.15036	-0.10	0.920	-.3099347	.279589
state_38	-.1091308	.1483587	-0.74	0.462	-.39997	.1817085
state_39	-.0806922	.1456463	-0.55	0.580	-.3662157	.2048313
state_40	-.1229746	.1592183	-0.77	0.440	-.4350978	.1891486
state_41	-.0047064	.1483494	-0.03	0.975	-.2955279	.2861151
state_42	.0044938	.1604274	0.03	0.978	-.3099992	.3189868
state_43	-.0009327	.1474286	-0.01	0.995	-.2899495	.288084
state_44	-.0171917	.1453409	-0.12	0.906	-.3021167	.2677333
state_45	.0128714	.1503539	0.09	0.932	-.281878	.3076208
state_46	-.0364566	.1467464	-0.25	0.804	-.3241358	.2512226
state_47	-.2259235	.1624484	-1.39	0.164	-.5443777	.0925308
state_48	-.133635	.1467729	-0.91	0.363	-.4213662	.1540962
state_49	-.0731084	.1472463	-0.50	0.620	-.3617677	.215551
state_50	-.1065127	.1512541	-0.70	0.481	-.4030268	.1900013
state_51	.0193391	.1661122	0.12	0.907	-.3062967	.3449749
_cons	.6381649	.1561839	4.09	0.000	.3319883	.9443415

```
33 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

34 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**-.0897892**	**.0388145**	**-2.31**	**0.021**	**-.1658752**	**-.0137032**

```
35 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.0261768**	**.0273364**	**0.96**	**0.338**	**-.0274092**	**.0797629**

```
36 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.1421428**	**.0172283**	**8.25**	**0.000**	**.1083712**	**.1759144**

```
37 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.2581088**	**.0124567**	**20.72**	**0.000**	**.2336905**	**.2825271**

```
38 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.3740748**	**.0179541**	**20.84**	**0.000**	**.3388803**	**.4092693**

```
39 . mi estimate: regress ideo c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.race##c
   > lengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

   Multiple-imputation estimates                   Imputations       =         25
   Linear regression                               Number of obs     =      8,527
                                                   Average RVI       =     0.0027
                                                   Largest FMI       =     0.0484
                                                   Complete DF       =       8453
   DF adjustment:   Small sample                   DF:     min       =   4,518.13
                                                           avg       =   7,416.51
                                                           max       =   8,447.00
   Model F test:       Equal FMI                   F(  73, 8450.7)   =      23.39
   Within VCE type:          OLS                   Prob > F          =     0.0000
```

ideo	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.0280487	.028748	0.98	0.329	-.0283044	.0844019
polengage_scale	-.2398098	.0566824	-4.23	0.000	-.3509212	-.1286983
c.auth_scale#c.polengage_scale	.3289449	.0368012	8.94	0.000	.2568056	.4010843
age	.1151375	.041457	2.78	0.005	.0338709	.196404
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.0223222	.0564633	-0.40	0.693	-.1330047	.0883603
1.male	-.0176556	.0181373	-0.97	0.330	-.0532091	.017898
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.0595717	.0235903	2.53	0.012	.0133288	.1058146
race						
1	-.011592	.0332114	-0.35	0.727	-.0766946	.0535105
2	-.097485	.0388698	-2.51	0.012	-.1736795	-.0212906
3	-.0932601	.0394632	-2.36	0.018	-.1706177	-.0159025
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.0459919	.0451832	1.02	0.309	-.0425782	.134562
2	-.0454849	.0548363	-0.83	0.407	-.1529775	.0620077
3	.0786306	.0556709	1.41	0.158	-.0304981	.1877593
education	.008491	.0341323	0.25	0.804	-.0584169	.0753989
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.1057883	.0442961	-2.39	0.017	-.1926201	-.0189566
income	-.0117423	.0340451	-0.34	0.730	-.0784822	.0549976
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.0791018	.0451796	1.75	0.080	-.0094722	.1676758
sample_num						
2	-.0144843	.0161185	-0.90	0.369	-.0460807	.0171121
3	.0052193	.0128327	0.41	0.684	-.019936	.0303745
4	-.0159936	.0115673	-1.38	0.167	-.0386684	.0066812
5	.015518	.0151463	1.02	0.306	-.0141725	.0452085
6	.0271713	.0111203	2.44	0.015	.0053727	.04897
state_1	-.2006559	.1188426	-1.69	0.091	-.4336213	.0323094
state_2	-.0305752	.1090317	-0.28	0.779	-.2443099	.1831596
state_3	-.0116956	.1098227	-0.11	0.915	-.2269807	.2035896
state_4	-.0497474	.1075226	-0.46	0.644	-.2605241	.1610294
state_5	-.1110219	.1065268	-1.04	0.297	-.3198468	.0978029
state_6	-.0679137	.1082042	-0.63	0.530	-.2800261	.1441987
state_7	-.0929527	.1105446	-0.84	0.400	-.3096531	.1237478
state_8	-.1035887	.1174033	-0.88	0.378	-.3337327	.1265553
state_9	-.1411599	.1159398	-1.22	0.223	-.3684353	.0861155
state_10	-.0479877	.1066128	-0.45	0.653	-.2569812	.1610057
state_11	-.045645	.1072115	-0.43	0.670	-.2558122	.1645221
state_12	-.0888591	.1148529	-0.77	0.439	-.314004	.1362858
state_13	-.0671895	.1095677	-0.61	0.540	-.2819749	.1475958
state_14	-.0305695	.1120073	-0.27	0.785	-.2501368	.1889977
state_15	-.0867134	.107162	-0.81	0.418	-.2967835	.1233566
state_16	-.0401041	.1080627	-0.37	0.711	-.2519394	.1717312
state_17	-.0579914	.1098327	-0.53	0.598	-.2732961	.1573133
state_18	-.1017149	.1090536	-0.93	0.351	-.3154927	.1120629
state_19	-.0610607	.1097374	-0.56	0.578	-.2761787	.1540573

state_20	-.1410063	.1081121	-1.30	0.192	-.3529385	.0709259
state_21	-.0984538	.1081767	-0.91	0.363	-.3105127	.1136051
state_22	-.0842815	.1130763	-0.75	0.456	-.305944	.137381
state_23	-.0875363	.1072897	-0.82	0.415	-.2978565	.1227838
state_24	-.0145053	.1087876	-0.13	0.894	-.2277616	.1987511
state_25	-.031634	.1079804	-0.29	0.770	-.2433082	.1800403
state_26	-.0577487	.111412	-0.52	0.604	-.2761491	.1606518
state_27	-.0180458	.1154368	-0.16	0.876	-.2443356	.208244
state_28	-.0756957	.1074339	-0.70	0.481	-.2862987	.1349073
state_29	-.1883443	.1250582	-1.51	0.132	-.4334931	.0568044
state_30	-.0065124	.1126617	-0.06	0.954	-.2273624	.2143375
state_31	-.0847984	.1139726	-0.74	0.457	-.3082178	.138621
state_32	-.0843538	.1075197	-0.78	0.433	-.2951249	.1264173
state_33	-.0185001	.1128652	-0.16	0.870	-.2397489	.2027486
state_34	-.0685878	.1096543	-0.63	0.532	-.2835427	.1463671
state_35	-.1193221	.1066875	-1.12	0.263	-.328462	.0898178
state_36	-.0910022	.1071794	-0.85	0.396	-.3011063	.119102
state_37	-.0435745	.110231	-0.40	0.693	-.2596604	.1725113
state_38	-.0969046	.1089182	-0.89	0.374	-.3104169	.1166078
state_39	-.0855298	.1068882	-0.80	0.424	-.2950631	.1240035
state_40	-.1325369	.1170155	-1.13	0.257	-.3619209	.0968471
state_41	-.0285711	.1088775	-0.26	0.793	-.2420038	.1848616
state_42	-.0044223	.1178903	-0.04	0.970	-.235521	.2266764
state_43	-.0289808	.1082009	-0.27	0.789	-.2410873	.1831256
state_44	-.0333436	.1066519	-0.31	0.755	-.2424136	.1757264
state_45	-.0694229	.1104013	-0.63	0.529	-.2858422	.1469963
state_46	-.0533721	.1077176	-0.50	0.620	-.2645309	.1577868
state_47	-.2327999	.1193921	-1.95	0.051	-.4668424	.0012425
state_48	-.0967706	.1077094	-0.90	0.369	-.3079134	.1143723
state_49	-.085414	.1080516	-0.79	0.429	-.2972278	.1263999
state_50	-.1252092	.1110864	-1.13	0.260	-.3429716	.0925531
state_51	-.0169644	.1221145	-0.14	0.890	-.2563434	.2224146
_cons	.5804899	.1146811	5.06	0.000	.3556813	.8052985

```
40 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

41 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0280487	.028748	0.98	0.329	-.0283044	.0844019

```
42 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.110285	.0202464	5.45	0.000	.070597	.1499729

```
43 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.1925212**	**.0127545**	**15.09**	**0.000**	**.1675193**	**.2175231**

```
44 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.2747574**	**.0092047**	**29.85**	**0.000**	**.256714**	**.2928009**

```
45 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.3569936**	**.013269**	**26.90**	**0.000**	**.3309831**	**.3830042**

```
46 . mi estimate: regress leftright c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.ra
   > #c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      8,601
                                                Average RVI       =     0.0033
                                                Largest FMI       =     0.0509
                                                Complete DF       =       8527
DF adjustment:   Small sample                   DF:     min       =   4,338.98
                                                        avg       =   6,519.88
                                                        max       =   8,521.33
Model F test:       Equal FMI                   F(  73, 8524.5)   =      25.86
Within VCE type:          OLS                   Prob > F          =     0.0000
```

leftright	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	**-.0312308**	**.0304536**	**-1.03**	**0.305**	**-.0909273**	**.0284657**
polengage_scale	**-.300804**	**.0602501**	**-4.99**	**0.000**	**-.4189089**	**-.1826991**
c.auth_scale#c.polengage_scale	**.3982943**	**.0390144**	**10.21**	**0.000**	**.3218167**	**.4747719**
age	**.0540415**	**.0438735**	**1.23**	**0.218**	**-.0319616**	**.1400447**
polengage_scale	**0**	(omitted)				
c.age#c.polengage_scale	**.0513091**	**.0597984**	**0.86**	**0.391**	**-.0659107**	**.1685289**
1.male	**-.0089013**	**.0192307**	**-0.46**	**0.643**	**-.0465982**	**.0287955**
polengage_scale	**0**	(omitted)				
male#c.polengage_scale						
1	**.0599401**	**.0250304**	**2.39**	**0.017**	**.0108743**	**.1090058**
race						
1	**.032963**	**.0353105**	**0.93**	**0.351**	**-.0362541**	**.1021801**
2	**-.16961**	**.0413248**	**-4.10**	**0.000**	**-.2506166**	**-.0886035**
3	**-.1061815**	**.0418307**	**-2.54**	**0.011**	**-.1881799**	**-.0241832**

polengage_scale	**0**	(omitted)				
race#c.polengage_scale						
1	**-.0104585**	**.0480525**	**-0.22**	**0.828**	**-.104653**	**.0837361**
2	**-.0624737**	**.0583257**	**-1.07**	**0.284**	**-.1768064**	**.0518589**
3	**.054042**	**.0590339**	**0.92**	**0.360**	**-.0616789**	**.1697628**
education	**-.0243318**	**.0362402**	**-0.67**	**0.502**	**-.0953715**	**.046708**
polengage_scale	**0**	(omitted)				
c.education#c.polengage_scale	**-.0665285**	**.0470602**	**-1.41**	**0.157**	**-.1587785**	**.0257216**
income	**-.003128**	**.0361198**	**-0.09**	**0.931**	**-.0739354**	**.0676794**
polengage_scale	**0**	(omitted)				
c.income#c.polengage_scale	**.0806143**	**.0479979**	**1.68**	**0.093**	**-.0134862**	**.1747148**
sample_num						
2	**-.0069477**	**.0168792**	**-0.41**	**0.681**	**-.0400353**	**.0261399**
3	**.017442**	**.0136101**	**1.28**	**0.200**	**-.0092371**	**.0441211**
4	**-.0294599**	**.0122603**	**-2.40**	**0.016**	**-.0534931**	**-.0054268**
5	**.0196776**	**.0158115**	**1.24**	**0.213**	**-.0113167**	**.0506719**
6	**.0162185**	**.0117928**	**1.38**	**0.169**	**-.0068983**	**.0393354**
state_1	**-.2141341**	**.1272123**	**-1.68**	**0.092**	**-.4635134**	**.0352453**
state_2	**-.0092519**	**.1168019**	**-0.08**	**0.937**	**-.2382277**	**.219724**
state_3	**-.0136858**	**.1175724**	**-0.12**	**0.907**	**-.2441715**	**.2167999**
state_4	**-.0321308**	**.1152355**	**-0.28**	**0.780**	**-.2580367**	**.193775**
state_5	**-.1167609**	**.1141845**	**-1.02**	**0.307**	**-.3406068**	**.1070851**
state_6	**-.0523006**	**.1159511**	**-0.45**	**0.652**	**-.2796082**	**.1750069**
state_7	**-.1084905**	**.1184459**	**-0.92**	**0.360**	**-.3406889**	**.1237079**
state_8	**-.1205768**	**.125708**	**-0.96**	**0.338**	**-.3670078**	**.1258541**
state_9	**-.1405319**	**.1241522**	**-1.13**	**0.258**	**-.3839135**	**.1028497**
state_10	**-.0516524**	**.1142759**	**-0.45**	**0.651**	**-.2756774**	**.1723726**
state_11	**-.037139**	**.1149083**	**-0.32**	**0.747**	**-.2624037**	**.1881256**
state_12	**-.0857521**	**.1229962**	**-0.70**	**0.486**	**-.3268681**	**.1553639**
state_13	**-.0742933**	**.1172758**	**-0.63**	**0.526**	**-.304198**	**.1556114**
state_14	**-.0204828**	**.1199791**	**-0.17**	**0.864**	**-.2556852**	**.2147196**
state_15	**-.0782785**	**.114857**	**-0.68**	**0.496**	**-.3034426**	**.1468857**
state_16	**-.0324899**	**.1157973**	**-0.28**	**0.779**	**-.2594967**	**.1945169**
state_17	**-.0425693**	**.1175919**	**-0.36**	**0.717**	**-.2730933**	**.1879547**
state_18	**-.0930935**	**.1168372**	**-0.80**	**0.426**	**-.3221387**	**.1359517**
state_19	**-.0288017**	**.11755**	**-0.25**	**0.806**	**-.2592436**	**.2016402**
state_20	**-.1455134**	**.115862**	**-1.26**	**0.209**	**-.372647**	**.0816202**
state_21	**-.1137325**	**.1159057**	**-0.98**	**0.327**	**-.340952**	**.113487**
state_22	**-.0847511**	**.1211116**	**-0.70**	**0.484**	**-.3221733**	**.1526711**
state_23	**-.0866305**	**.1149742**	**-0.75**	**0.451**	**-.3120242**	**.1387631**
state_24	**-.0185325**	**.1165803**	**-0.16**	**0.874**	**-.247074**	**.2100091**
state_25	**-.0380124**	**.11569**	**-0.33**	**0.742**	**-.2648093**	**.1887844**
state_26	**-.0275084**	**.1192508**	**-0.23**	**0.818**	**-.2612836**	**.2062667**
state_27	**-.0424253**	**.1236005**	**-0.34**	**0.731**	**-.2847258**	**.1998753**
state_28	**-.05958**	**.1151443**	**-0.52**	**0.605**	**-.2853073**	**.1661473**
state_29	**-.1740578**	**.1338001**	**-1.30**	**0.193**	**-.4363491**	**.0882335**
state_30	**.0143634**	**.120675**	**0.12**	**0.905**	**-.2222032**	**.2509301**
state_31	**-.0903652**	**.1218576**	**-0.74**	**0.458**	**-.3292498**	**.1485194**
state_32	**-.0815111**	**.1152237**	**-0.71**	**0.479**	**-.3073939**	**.1443716**
state_33	**.0348108**	**.1207144**	**0.29**	**0.773**	**-.2018325**	**.2714542**
state_34	**-.0693854**	**.1175008**	**-0.59**	**0.555**	**-.2997307**	**.16096**
state_35	**-.130494**	**.1143524**	**-1.14**	**0.254**	**-.3546692**	**.0936812**
state_36	**-.0839457**	**.1148666**	**-0.73**	**0.465**	**-.3091285**	**.1412371**
state_37	**-.0242852**	**.1180467**	**-0.21**	**0.837**	**-.255701**	**.2071306**
state_38	**-.1017567**	**.116722**	**-0.87**	**0.383**	**-.3305759**	**.1270624**
state_39	**-.0821721**	**.1145755**	**-0.72**	**0.473**	**-.3067845**	**.1424404**
state_40	**-.1264852**	**.1252754**	**-1.01**	**0.313**	**-.3720684**	**.119098**
state_41	**-.0130899**	**.1166215**	**-0.11**	**0.911**	**-.2417125**	**.2155327**
state_42	**.0015657**	**.1262122**	**0.01**	**0.990**	**-.2458535**	**.2489849**

state_43	-.0132235	.1159318	-0.11	0.909	-.2404943	.2140473
state_44	-.0235964	.114322	-0.21	0.836	-.247712	.2005192
state_45	-.0272178	.1182867	-0.23	0.818	-.2591033	.2046676
state_46	-.0446251	.1154252	-0.39	0.699	-.2709025	.1816523
state_47	-.227721	.1278018	-1.78	0.075	-.4782559	.0228139
state_48	-.1170923	.1154283	-1.01	0.310	-.3433757	.1091912
state_49	-.0756964	.1158098	-0.65	0.513	-.302728	.1513351
state_50	-.1146964	.1190128	-0.96	0.335	-.3480055	.1186127
state_51	.0026099	.1307027	0.02	0.984	-.2536112	.2588311
_cons	.6094557	.1227625	4.96	0.000	.3687972	.8501142

```
47 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

48 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0312308	.0304536	-1.03	0.305	-.0909273	.0284657

```
49 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.0683428	.0214444	3.19	0.001	.0263066	.1103789

```
50 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.1679163	.0135116	12.43	0.000	.1414304	.1944023

```
51 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.2674899	.0097739	27.37	0.000	.2483307	.2866491

```
52 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.3670635	.0140982	26.04	0.000	.3394275	.3946994

```
53 . mi estimate: regress pid c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.race##c.
> engage_scale c.income##c.polengage_scale i.sample_num

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =        940
                                              Average RVI       =     0.0131
                                              Largest FMI       =     0.0386
                                              Complete DF       =        921
DF adjustment:   Small sample                 DF:     min       =     838.17
                                                      avg       =     893.63
                                                      max       =     917.73
Model F test:       Equal FMI                 F(  18,  918.6)   =       9.53
Within VCE type:          OLS                 Prob > F          =     0.0000
```

pid	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.3359037	.2557065	-1.31	0.189	-.837741	.1659336
polengage_scale	-.5358001	.3106625	-1.72	0.085	-1.1455	.0738994
c.latentnsc2#c.polengage_scale	1.153284	.3267252	3.53	0.000	.5120686	1.794499
age	.0503458	.2190174	0.23	0.818	-.379489	.4801805
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.0647508	.2842623	-0.23	0.820	-.6226324	.4931308
1.male	.1495526	.0927502	1.61	0.107	-.032475	.3315803
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.1662827	.1226442	-1.36	0.175	-.4069792	.0744137
race						
1	.007105	.1791846	0.04	0.968	-.3445636	.3587735
2	-.2476469	.2303752	-1.07	0.283	-.699803	.2045092
3	-.2664022	.2095327	-1.27	0.204	-.6776398	.1448354
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.1125981	.2494768	0.45	0.652	-.3770286	.6022248
2	.0349707	.3294107	0.11	0.915	-.6115721	.6815136
3	.2982288	.2951751	1.01	0.313	-.2810964	.8775541
education	.3110349	.2013189	1.54	0.123	-.0840831	.7061529
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.4840501	.27277	-1.77	0.076	-1.0194	.0512995
income	-.1090826	.1616165	-0.67	0.500	-.4263032	.208138
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.2157542	.2125142	1.02	0.310	-.201368	.6328765
5.sample_num	.0284955	.0278678	1.02	0.307	-.0261966	.0831877
_cons	.5546568	.2238873	2.48	0.013	.1152631	.9940505

```
54 . quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

55 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**-.244202**	**.1858986**	**-1.31**	**0.189**	**-.609036**	**.120632**

```
56 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**-.0345926**	**.1289708**	**-0.27**	**0.789**	**-.2877033**	**.2185181**

```
57 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.1750168**	**.0759022**	**2.31**	**0.021**	**.0260554**	**.3239781**

```
58 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.3846261**	**.0440608**	**8.73**	**0.000**	**.2981549**	**.4710973**

```
59 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.5942355**	**.0719311**	**8.26**	**0.000**	**.4530677**	**.7354033**

```
60 . mi estimate: regress ideo c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.race##c
   > lengage_scale c.income##c.polengage_scale i.sample_num

   Multiple-imputation estimates                   Imputations       =         25
   Linear regression                               Number of obs     =        980
                                                   Average RVI       =     0.0170
                                                   Largest FMI       =     0.0330
                                                   Complete DF       =        961
   DF adjustment:   Small sample                   DF:     min       =     890.49
                                                           avg       =     933.78
                                                           max       =     956.96
   Model F test:       Equal FMI                   F(  18,  958.3)   =      15.73
   Within VCE type:          OLS                   Prob > F          =     0.0000
```

ideo	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.3719479	.1862416	-2.00	0.046	-.7374369	-.0064589
polengage_scale	-.4914408	.2201674	-2.23	0.026	-.9235197	-.0593619
c.latentnsc2#c.polengage_scale	1.202915	.2380838	5.05	0.000	.7356886	1.670142
age	.2592878	.159051	1.63	0.103	-.0528443	.5714199
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.1440035	.2071818	-0.70	0.487	-.5505926	.2625856
1.male	.0835514	.0658836	1.27	0.205	-.0457419	.2128446
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0773034	.0873851	-0.88	0.377	-.2487931	.0941863
race						
1	.0360923	.1204516	0.30	0.765	-.2002943	.272479
2	-.0017728	.156576	-0.01	0.991	-.3090577	.3055122
3	-.1170993	.1450782	-0.81	0.420	-.4018195	.1676209
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.0619513	.1713191	0.36	0.718	-.2742663	.3981689
2	-.0575639	.2267915	-0.25	0.800	-.5026547	.3875269
3	.1799472	.2075111	0.87	0.386	-.2272979	.5871924
education	.408728	.1414078	2.89	0.004	.1312139	.6862421
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.5667216	.1917724	-2.96	0.003	-.9430929	-.1903504
income	-.0477106	.1150748	-0.41	0.679	-.2735564	.1781353
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.1821715	.1509794	1.21	0.228	-.1141454	.4784884
5.sample_num	.0340874	.0205942	1.66	0.098	-.0063278	.0745025
_cons	.4412187	.1563439	2.82	0.005	.1343982	.7480392

```
61 . quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

62 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.2704062	.1353976	-2.00	0.046	-.5361152	-.0046971

```
63 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0517763	.0939689	-0.55	0.582	-.2361842	.1326317

```
64 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.1668536	.0554319	3.01	0.003	.0580721	.275635

```
65 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.3854835	.0325593	11.84	0.000	.3215879	.449379

```
66 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.6041133	.0528433	11.43	0.000	.5004118	.7078149

```
67 . mi estimate: regress leftright c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.ra
   > #c.polengage_scale c.income##c.polengage_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      1,040
                                                Average RVI       =     0.0138
                                                Largest FMI       =     0.0326
                                                Complete DF       =       1021
DF adjustment:   Small sample                   DF:     min       =     945.19
                                                        avg       =     993.73
                                                        max       =   1,017.43
Model F test:       Equal FMI                   F(  18, 1018.4)   =      14.61
Within VCE type:          OLS                   Prob > F          =     0.0000
```

leftright	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.3388198	.1946925	-1.74	0.082	-.7208646	.0432251
polengage_scale	-.5328693	.2353844	-2.26	0.024	-.9947723	-.0709662
c.latentnsc2#c.polengage_scale	1.178909	.2501769	4.71	0.000	.6879868	1.66983
age	.2060039	.167799	1.23	0.220	-.1232693	.5352771
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.1622797	.2188762	-0.74	0.459	-.5917825	.2672231

1.male	.1155501	.0696435	1.66	0.097	-.0211115	.2522117
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.1170729	.0925933	-1.26	0.206	-.2987692	.0646234
race						
1	-.0122871	.1294081	-0.09	0.924	-.26623	.2416557
2	-.1520452	.1660096	-0.92	0.360	-.4778228	.1737324
3	-.2252756	.1551546	-1.45	0.147	-.5297499	.0791986
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.1232412	.1840963	0.67	0.503	-.2380213	.4845037
2	.0238927	.2413116	0.10	0.921	-.4496648	.4974502
3	.2699128	.2222316	1.21	0.225	-.1661912	.7060168
education	.3479208	.1499037	2.32	0.020	.053757	.6420846
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.5070766	.2037819	-2.49	0.013	-.9069748	-.1071784
income	-.0917507	.1220045	-0.75	0.452	-.3311788	.1476773
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.2210955	.1606731	1.38	0.169	-.0942217	.5364127
5.sample_num	.0303939	.0215617	1.41	0.159	-.0119167	.0727045
_cons	.5073983	.1671382	3.04	0.002	.1794204	.8353762

```
68 . quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

69 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.246322	.1415415	-1.74	0.082	-.5240674	.0314235

```
70 .     lincom _b[7._at]-_b[2._at]

 ( 1)  - 2._at + 7._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0320553	.0980519	-0.33	0.744	-.2244617	.160351

```
71 .     lincom _b[8._at]-_b[3._at]

 ( 1)  - 3._at + 8._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.1822113	.0577006	3.16	0.002	.068986	.2954366

```
72 .     lincom _b[9._at]-_b[4._at]

 ( 1)  - 4._at + 9._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.3964779**	**.034344**	**11.54**	**0.000**	**.3290851**	**.4638707**

```
73 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.6107446**	**.0562551**	**10.86**	**0.000**	**.5003557**	**.7211334**

```
74 .
75 .
end of do-file

76 .
```

```
21 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000005.tmp"

22 .
23 . ***** SM4 - ALTERNATIVE MODEL SPECIFICATIONS *****
24 .
25 . *Concern, Authoritarianism
26 . mi estimate: regress dv_concern c.auth_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      5,485
                                                Average RVI       =     0.0000
                                                Largest FMI       =     0.0000
                                                Complete DF       =       5480
DF adjustment:   Small sample                   DF:     min       =   5,478.00
                                                        avg       =   5,478.00
                                                        max       =   5,478.00
Model F test:       Equal FMI                   F(   4, 5478.0)   =      66.83
Within VCE type:          OLS                   Prob > F          =     0.0000

------------------------------------------------------------------------------
  dv_concern | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
  auth_scale |  -.1637222   .0130199   -12.57   0.000    -.1892464    -.138198

  sample_num |
           2 |  -.1328466   .0185921    -7.15   0.000    -.1692944   -.0963988
           5 |  -.1724373   .0181061    -9.52   0.000    -.2079325   -.1369421
           6 |  -.0539078   .0124855    -4.32   0.000    -.0783842   -.0294313

       _cons |    .727447   .0128384    56.66   0.000     .7022786    .7526155
------------------------------------------------------------------------------
```

```
27 . mi estimate: regress dv_concern c.auth_scale c.age i.male i.race c.education c.income i.sample_num state_1-state_51 po

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      5,484
                                                Average RVI       =     0.0030
                                                Largest FMI       =     0.1342
                                                Complete DF       =       5420
DF adjustment:   Small sample                   DF:     min       =   1,054.20
                                                        avg       =   5,255.66
                                                        max       =   5,414.53
Model F test:       Equal FMI                   F(  63, 5417.8)   =       7.92
Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.1701995	.013967	-12.19	0.000	-.1975805	-.1428186
age	.0053203	.0230869	0.23	0.818	-.0399393	.0505799
1.male	-.0586108	.0089821	-6.53	0.000	-.0762193	-.0410022
race						
1	-.0512203	.0188149	-2.72	0.007	-.0881051	-.0143355
2	.0779661	.0232196	3.36	0.001	.0324463	.123486
3	.0062868	.0244088	0.26	0.797	-.0415644	.054138
education	.0361902	.0170414	2.12	0.034	.0027815	.0695989
income	-.0679963	.0214029	-3.18	0.002	-.1099935	-.0259991
sample_num						
2	-.1648529	.0208143	-7.92	0.000	-.2056591	-.1240467
5	-.1911964	.0188781	-10.13	0.000	-.2282053	-.1541875
6	-.0712355	.0144864	-4.92	0.000	-.0996362	-.0428349
state_1	-.0110635	.1607927	-0.07	0.945	-.3262824	.3041554
state_2	-.0426713	.1468344	-0.29	0.771	-.3305265	.2451838
state_3	.0643435	.1474766	0.44	0.663	-.2247708	.3534578
state_4	-.0957182	.1430105	-0.67	0.503	-.3760772	.1846408
state_5	.0113251	.1411341	0.08	0.936	-.2653552	.2880055
state_6	.00719	.1444384	0.05	0.960	-.275968	.2903481
state_7	.0273698	.1489005	0.18	0.854	-.2645359	.3192755
state_8	.1586198	.1714294	0.93	0.355	-.1774512	.4946909
state_9	.0711053	.1558969	0.46	0.648	-.234516	.3767266
state_10	-.002114	.1412448	-0.01	0.988	-.2790113	.2747834
state_11	-.0119374	.1422858	-0.08	0.933	-.2908756	.2670008
state_12	.0507222	.1586596	0.32	0.749	-.260315	.3617594
state_13	.0416799	.1450061	0.29	0.774	-.2425911	.325951
state_14	-.0207099	.1485683	-0.14	0.889	-.3119643	.2705445
state_15	.0216841	.1420702	0.15	0.879	-.2568315	.3001997
state_16	-.0182775	.1438114	-0.13	0.899	-.3002064	.2636515
state_17	.0210237	.1464569	0.14	0.886	-.2660915	.3081389
state_18	.0518353	.1460587	0.35	0.723	-.2344992	.3381699
state_19	-.0460226	.1485693	-0.31	0.757	-.3372789	.2452336
state_20	.0075596	.1439839	0.05	0.958	-.2747075	.2898267
state_21	.0834843	.143778	0.58	0.562	-.1983792	.3653479
state_22	.0749322	.1518205	0.49	0.622	-.2226977	.3725622
state_23	-.0405105	.1422692	-0.28	0.776	-.3194162	.2383952
state_24	-.0376337	.1448694	-0.26	0.795	-.3216369	.2463694
state_25	-.0268891	.1437214	-0.19	0.852	-.3086417	.2548634
state_26	.0507733	.1488238	0.34	0.733	-.2409819	.3425285
state_27	-.0622902	.1574988	-0.40	0.692	-.3710518	.2464715
state_28	-.0150825	.1426821	-0.11	0.916	-.2947977	.2646326
state_29	-.1834575	.1975078	-0.93	0.353	-.5706526	.2037376
state_30	-.0074814	.1530474	-0.05	0.961	-.3075165	.2925538
state_31	-.0009871	.1530965	-0.01	0.995	-.3011184	.2991442
state_32	.0211563	.1428675	0.15	0.882	-.2589223	.3012348
state_33	-.0683507	.150692	-0.45	0.650	-.3637682	.2270669
state_34	-.0157189	.147249	-0.11	0.915	-.3043869	.2729491

state_35	.051305	.1415621	0.36	0.717	-.2262144	.3288245
state_36	.03476	.1422594	0.24	0.807	-.2441264	.3136464
state_37	.0868689	.1492421	0.58	0.561	-.2057063	.3794442
state_38	-.0552503	.1454298	-0.38	0.704	-.340352	.2298513
state_39	.0169791	.1417007	0.12	0.905	-.260812	.2947702
state_40	.0926559	.1605392	0.58	0.564	-.2220662	.4073779
state_41	.0019391	.1456813	0.01	0.989	-.2836557	.2875338
state_42	.0381774	.1560708	0.24	0.807	-.2677847	.3441396
state_43	-.003671	.1440397	-0.03	0.980	-.2860476	.2787056
state_44	-.0205676	.1413514	-0.15	0.884	-.2976739	.2565387
state_45	-.0379181	.1473514	-0.26	0.797	-.326787	.2509507
state_46	-.007768	.143173	-0.05	0.957	-.2884453	.2729094
state_47	.0919819	.1648717	0.56	0.577	-.2312335	.4151973
state_48	-.0560875	.1429251	-0.39	0.695	-.3362789	.224104
state_49	-.0020638	.1433557	-0.01	0.989	-.2830995	.2789719
state_50	.0557449	.148993	0.37	0.708	-.2363421	.3478319
state_51	-.0336819	.1714798	-0.20	0.844	-.3698519	.302488
polengage_scale	.0496103	.0210514	2.36	0.018	.0083411	.0908795
_cons	.7738325	.1437362	5.38	0.000	.4920506	1.055614

28 . mi estimate: regress dv_concern c.auth_scale##c.polengage_scale i.sample_num

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 5,484
Average RVI = 0.0000
Largest FMI = 0.0000
Complete DF = 5477
DF adjustment: **Small sample** — DF: min = 5,475.00
avg = 5,475.00
max = 5,475.00
Model F test: **Equal FMI** — F(6, 5475.0) = 57.28
Within VCE type: **OLS** — Prob > F = 0.0000

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.2149341	.0473207	4.54	0.000	.1221667	.3077014
polengage_scale	.2479917	.0374253	6.63	0.000	.1746232	.3213601
c.auth_scale#c.polengage_scale	-.4834782	.0572729	-8.44	0.000	-.5957559	-.3712006
sample_num						
2	-.1220212	.0185465	-6.58	0.000	-.1583796	-.0856627
5	-.1618524	.018128	-8.93	0.000	-.1973905	-.1263142
6	-.052194	.0128268	-4.07	0.000	-.0773396	-.0270485
_cons	.5205981	.0325921	15.97	0.000	.4567046	.5844916

29 . mi estimate: regress dv_concern c.auth_scale##c.polengage_scale c.age i.male i.race c.education c.income i.sample_num

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 5,484
Average RVI = 0.0028
Largest FMI = 0.1251
Complete DF = 5419
DF adjustment: **Small sample** — DF: min = 1,175.42
avg = 5,262.14
max = 5,414.76
Model F test: **Equal FMI** — F(64, 5416.8) = 8.92
Within VCE type: **OLS** — Prob > F = 0.0000

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.1939798	.0470373	4.12	0.000	.1017677	.2861919
polengage_scale	.3078089	.0381126	8.08	0.000	.2330927	.382525
c.auth_scale#c.polengage_scale	-.4602681	.0567976	-8.10	0.000	-.5716142	-.3489221
age	.0016451	.0229549	0.07	0.943	-.0433558	.046646
1.male	-.0588364	.0089282	-6.59	0.000	-.0763394	-.0413335
race						
1	-.0539943	.0187076	-2.89	0.004	-.0906688	-.0173199
2	.0706553	.0231019	3.06	0.002	.0253663	.1159444
3	.0066938	.024268	0.28	0.783	-.0408813	.0542689
education	.0314179	.0169415	1.85	0.064	-.0017949	.0646306
income	-.0677464	.0211677	-3.20	0.001	-.1092772	-.0262157
sample_num						
2	-.1550034	.0207084	-7.49	0.000	-.1956016	-.1144051
5	-.1831961	.0187874	-9.75	0.000	-.220027	-.1463651
6	-.0731634	.0143854	-5.09	0.000	-.1013659	-.0449609
state_1	.001283	.1598542	0.01	0.994	-.3120961	.3146621
state_2	-.0339967	.1459744	-0.23	0.816	-.320166	.2521726
state_3	.0696297	.1466114	0.47	0.635	-.2177884	.3570477
state_4	-.0870449	.1421756	-0.61	0.540	-.365767	.1916773
state_5	.01462	.1403076	0.10	0.917	-.2604401	.2896802
state_6	.0084499	.1435906	0.06	0.953	-.2730463	.289946
state_7	.0289656	.1480278	0.20	0.845	-.2612292	.3191604
state_8	.1599571	.1704228	0.94	0.348	-.1741406	.4940549
state_9	.0752925	.154981	0.49	0.627	-.2285333	.3791183
state_10	.0041136	.1404182	0.03	0.977	-.2711634	.2793906
state_11	-.0027605	.1414557	-0.02	0.984	-.2800715	.2745505
state_12	.0693127	.1577454	0.44	0.660	-.2399323	.3785578
state_13	.0447777	.1441552	0.31	0.756	-.2378253	.3273807
state_14	-.0167338	.1476949	-0.11	0.910	-.3062759	.2728083
state_15	.0305739	.1412415	0.22	0.829	-.2463171	.3074649
state_16	-.0128512	.1429688	-0.09	0.928	-.2931282	.2674259
state_17	.0245826	.1455977	0.17	0.866	-.2608481	.3100134
state_18	.0553507	.1452012	0.38	0.703	-.2293029	.3400043
state_19	-.0417899	.1476979	-0.28	0.777	-.331338	.2477582
state_20	.0120395	.1431408	0.08	0.933	-.2685749	.2926538
state_21	.0897527	.1429378	0.63	0.530	-.1904636	.3699691
state_22	.0720908	.1509289	0.48	0.633	-.2237912	.3679728
state_23	-.0355297	.1414362	-0.25	0.802	-.3128024	.241743
state_24	-.0271842	.144026	-0.19	0.850	-.3095339	.2551656
state_25	-.0258658	.1428772	-0.18	0.856	-.3059634	.2542319
state_26	.0503366	.1479497	0.34	0.734	-.239705	.3403781
state_27	-.0513925	.1565774	-0.33	0.743	-.3583479	.2555629
state_28	-.0055776	.1418494	-0.04	0.969	-.2836602	.2725051
state_29	-.1781594	.1963453	-0.91	0.364	-.5630756	.2067568
state_30	-.0009498	.1521508	-0.01	0.995	-.2992272	.2973277
state_31	.0080731	.1522025	0.05	0.958	-.2903058	.306452
state_32	.0235257	.1420312	0.17	0.868	-.2549134	.3019647
state_33	-.0599092	.1498117	-0.40	0.689	-.353601	.2337826
state_34	-.00289	.146393	-0.02	0.984	-.28988	.2840999
state_35	.0543698	.1407329	0.39	0.699	-.2215241	.3302638
state_36	.0406781	.1414263	0.29	0.774	-.2365752	.3179314
state_37	.0882121	.1483641	0.59	0.552	-.202642	.3790663
state_38	-.0524915	.1445768	-0.36	0.717	-.335921	.2309381
state_39	.0246094	.140873	0.17	0.861	-.2515591	.300778
state_40	.1109083	.1596105	0.69	0.487	-.2019932	.4238097
state_41	.0041823	.1448268	0.03	0.977	-.2797373	.2881019
state_42	.0508897	.155161	0.33	0.743	-.253289	.3550684
state_43	-.0024295	.1431945	-0.02	0.986	-.2831492	.2782902

state_44	-.013347	.1405252	-0.09	0.924	-.2888337	.2621397
state_45	-.0314156	.1464891	-0.21	0.830	-.3185939	.2557627
state_46	.0005441	.1423373	0.00	0.997	-.278495	.2795831
state_47	.0956334	.1639044	0.58	0.560	-.2256856	.4169524
state_48	-.0507919	.142089	-0.36	0.721	-.3293443	.2277605
state_49	.0056325	.1425188	0.04	0.968	-.2737625	.2850275
state_50	.0513318	.1481169	0.35	0.729	-.2390377	.3417014
state_51	-.0318412	.1704705	-0.19	0.852	-.3660326	.3023502
_cons	.562476	.1452585	3.87	0.000	.2777099	.8472422

```
30 . mi estimate: regress dv_concern c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.r
> ##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51
```

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 5,484
Average RVI = 0.0037
Largest FMI = 0.0614
Complete DF = 5412
DF adjustment: **Small sample** — DF: min = 2,839.85
avg = 5,250.52
max = 5,406.85
Model F test: **Equal FMI** — F(71, 5409.7) = 8.30
Within VCE type: **OLS** — Prob > F = 0.0000

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.1698144	.0488236	3.48	0.001	.0741006	.2655282
polengage_scale	.2791857	.0989894	2.82	0.005	.0851264	.4732451
c.auth_scale#c.polengage_scale	-.4280247	.0594535	-7.20	0.000	-.5445775	-.3114719
age	.2139976	.074115	2.89	0.004	.0687024	.3592929
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.2840209	.0953901	-2.98	0.003	-.471024	-.0970178
1.male	.0042296	.032134	0.13	0.895	-.0587659	.0672252
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0799572	.0392459	-2.04	0.042	-.1568951	-.0030194
race						
1	-.1458162	.0588068	-2.48	0.013	-.2611012	-.0305311
2	.0011709	.0675823	0.02	0.986	-.1313178	.1336597
3	-.0224712	.0734298	-0.31	0.760	-.1664235	.1214811
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.1194261	.0746888	1.60	0.110	-.0269943	.2658464
2	.0907412	.0888589	1.02	0.307	-.0834585	.2649408
3	.0342578	.0953161	0.36	0.719	-.1526006	.2211163
education	-.0683349	.0572198	-1.19	0.232	-.1805094	.0438395
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.1289616	.0703315	1.83	0.067	-.0089172	.2668403
income	-.0681065	.0614003	-1.11	0.267	-.1884949	.0522818
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	.0020125	.0755249	0.03	0.979	-.1460767	.1501017
sample_num						

2	-.1528253	.0209469	-7.30	0.000	-.1938909	-.1117598
5	-.1811252	.018906	-9.58	0.000	-.2181887	-.1440616
6	-.0749138	.0146682	-5.11	0.000	-.1036705	-.0461572
state_1	.0054104	.159844	0.03	0.973	-.3079492	.3187699
state_2	-.0337456	.1459829	-0.23	0.817	-.319932	.2524409
state_3	.0731501	.1466206	0.50	0.618	-.2142864	.3605867
state_4	-.0834837	.1421906	-0.59	0.557	-.3622356	.1952682
state_5	.0179579	.140309	0.13	0.898	-.2571053	.2930211
state_6	.0082332	.1436126	0.06	0.954	-.2733064	.2897728
state_7	.0279018	.1480323	0.19	0.851	-.2623022	.3181059
state_8	.1703934	.1704803	1.00	0.318	-.1638174	.5046041
state_9	.083056	.1549788	0.54	0.592	-.2207658	.3868778
state_10	.0073475	.1404176	0.05	0.958	-.2679286	.2826236
state_11	-.0004104	.1414739	-0.00	0.998	-.2777573	.2769366
state_12	.0732503	.1578026	0.46	0.643	-.2361071	.3826077
state_13	.0502092	.1441543	0.35	0.728	-.2323924	.3328107
state_14	-.0111042	.1477027	-0.08	0.940	-.300662	.2784536
state_15	.0343601	.1412601	0.24	0.808	-.2425677	.3112879
state_16	-.0080554	.1429682	-0.06	0.955	-.2883318	.272221
state_17	.0295981	.1455978	0.20	0.839	-.2558334	.3150295
state_18	.0620919	.1452015	0.43	0.669	-.2225627	.3467466
state_19	-.0373625	.1477189	-0.25	0.800	-.3269522	.2522272
state_20	.015938	.1431355	0.11	0.911	-.2646662	.2965423
state_21	.0957577	.1429317	0.67	0.503	-.1844471	.3759624
state_22	.080559	.1509398	0.53	0.594	-.2153447	.3764627
state_23	-.0318926	.141428	-0.23	0.822	-.3091496	.2453645
state_24	-.024006	.1440517	-0.17	0.868	-.3064065	.2583944
state_25	-.0246559	.1428907	-0.17	0.863	-.3047803	.2554685
state_26	.0538343	.1479543	0.36	0.716	-.2362167	.3438852
state_27	-.0486302	.1565793	-0.31	0.756	-.3555896	.2583292
state_28	-.0012074	.1418671	-0.01	0.993	-.2793251	.2769104
state_29	-.1785343	.1963053	-0.91	0.363	-.5633722	.2063037
state_30	-.0015901	.1521536	-0.01	0.992	-.2998733	.2966931
state_31	.0119497	.1521974	0.08	0.937	-.2864194	.3103189
state_32	.0261768	.1420213	0.18	0.854	-.2522432	.3045968
state_33	-.0565228	.1498138	-0.38	0.706	-.3502191	.2371735
state_34	.0004768	.1463846	0.00	0.997	-.286497	.2874506
state_35	.0570415	.140732	0.41	0.685	-.2188509	.332934
state_36	.0446526	.1414324	0.32	0.752	-.2326129	.3219181
state_37	.0971401	.1483953	0.65	0.513	-.1937755	.3880558
state_38	-.0498003	.1445718	-0.34	0.731	-.3332203	.2336197
state_39	.0272937	.1408788	0.19	0.846	-.2488867	.3034742
state_40	.1023423	.159602	0.64	0.521	-.2105429	.4152275
state_41	.0082893	.1448058	0.06	0.954	-.2755895	.2921681
state_42	.0591517	.155159	0.38	0.703	-.2450233	.3633267
state_43	.0015716	.1432004	0.01	0.991	-.27916	.2823032
state_44	-.0098265	.1405358	-0.07	0.944	-.2853345	.2656814
state_45	-.0265837	.1465046	-0.18	0.856	-.3137928	.2606254
state_46	.006442	.1423456	0.05	0.964	-.2726138	.2854978
state_47	.1069835	.163897	0.65	0.514	-.2143214	.4282884
state_48	-.0487794	.142102	-0.34	0.731	-.3273576	.2297988
state_49	.0110171	.1425376	0.08	0.938	-.2684151	.2904493
state_50	.0555716	.1481115	0.38	0.708	-.2347877	.3459309
state_51	-.029108	.1704606	-0.17	0.864	-.3632802	.3050643
_cons	.5838799	.1611747	3.62	0.000	.2679114	.8998484

```
31 .
32 . *Behaviors, Authoritarianism
33 . mi estimate: regress dv_behavior c.auth_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      4,030
                                                Average RVI       =     0.0000
                                                Largest FMI       =     0.0000
                                                Complete DF       =       4025
DF adjustment:   Small sample                   DF:     min       =   4,023.00
                                                        avg       =   4,023.00
                                                        max       =   4,023.00
Model F test:       Equal FMI                   F(   4, 4023.0)   =     146.46
Within VCE type:          OLS                   Prob > F          =     0.0000

------------------------------------------------------------------------------
 dv_behavior | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
  auth_scale |  -.0379708   .0131934    -2.88   0.004    -.0638373   -.0121043
             |
  sample_num |
           3 |     .02802   .0136184     2.06   0.040     .0013204    .0547195
           4 |  -.1620291   .0124211   -13.04   0.000    -.1863813   -.1376769
           5 |   .0232471   .0155014     1.50   0.134    -.0071442    .0536385
             |
       _cons |    .840342   .0128637    65.33   0.000      .815122     .865562
------------------------------------------------------------------------------

34 . mi estimate: regress dv_behavior c.auth_scale c.age i.male i.race c.education c.income i.sample_num state_1-state_51 p

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      4,029
                                                Average RVI       =     0.0013
                                                Largest FMI       =     0.0214
                                                Complete DF       =       3965
DF adjustment:   Small sample                   DF:     min       =   3,611.37
                                                        avg       =   3,911.48
                                                        max       =   3,959.26
Model F test:       Equal FMI                   F(  63, 3963.0)   =      15.85
Within VCE type:          OLS                   Prob > F          =     0.0000

------------------------------------------------------------------------------
 dv_behavior | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
  auth_scale |  -.0371059   .0132352    -2.80   0.005    -.0630543   -.0111576
         age |   .0492223   .0182674     2.69   0.007     .0134069    .0850377
      1.male |  -.0238452   .0078346    -3.04   0.002    -.0392057   -.0084847
             |
        race |
           1 |  -.0206313   .0148315    -1.39   0.164    -.0497094    .0084469
           2 |   .0586049   .0183616     3.19   0.001     .0226057    .0946041
           3 |   .0312401   .0173707     1.80   0.072    -.0028164    .0652965
             |
   education |   .0584412    .015861     3.68   0.000     .0273446    .0895377
      income |   .0649205   .0144442     4.49   0.000     .0366009    .0932402
             |
  sample_num |
           3 |   .0115238   .0136263     0.85   0.398    -.0151915     .038239
           4 |  -.1559701   .0124543   -12.52   0.000    -.1803875   -.1315526
           5 |   .0065815   .0156958     0.42   0.675    -.0241912    .0373541
             |
     state_1 |   .1074409   .1461148     0.74   0.462    -.1790269    .3939087
     state_2 |   .0075781   .1206103     0.06   0.950    -.2288869     .244043
     state_3 |   .1018165   .1223557     0.83   0.405    -.1380704    .3417034
     state_4 |   .0579533   .1189445     0.49   0.626    -.1752458    .2911524
     state_5 |   .1329375   .1174505     1.13   0.258    -.0973325    .3632075
     state_6 |   .0168639    .120064     0.14   0.888    -.2185301    .2522579
```

state_7	.1241644	.1231838	1.01	0.314	-.117346	.3656748
state_8	.1602473	.1284477	1.25	0.212	-.0915833	.412078
state_9	.1164658	.1322435	0.88	0.379	-.1428067	.3757383
state_10	.0725547	.1176077	0.62	0.537	-.1580235	.303133
state_11	.0770215	.118398	0.65	0.515	-.1551061	.3091492
state_12	.2485858	.1289035	1.93	0.054	-.0041384	.50131
state_13	-.0140149	.1229858	-0.11	0.909	-.2551371	.2271073
state_14	.1452876	.1304564	1.11	0.265	-.1104811	.4010563
state_15	.0582342	.1185111	0.49	0.623	-.1741152	.2905836
state_16	.0379036	.1197913	0.32	0.752	-.1969557	.2727628
state_17	.0220279	.1221147	0.18	0.857	-.2173866	.2614424
state_18	.0883099	.1206965	0.73	0.464	-.148324	.3249439
state_19	.0475264	.1212293	0.39	0.695	-.1901521	.2852049
state_20	.1114418	.1200313	0.93	0.353	-.1238879	.3467715
state_21	.0902015	.1199438	0.75	0.452	-.1449567	.3253597
state_22	.1428923	.1305752	1.09	0.274	-.1131093	.3988938
state_23	.0678209	.1187388	0.57	0.568	-.1649748	.3006167
state_24	.0435072	.1211701	0.36	0.720	-.1940553	.2810698
state_25	.0325245	.1196406	0.27	0.786	-.2020392	.2670882
state_26	.0462217	.1246871	0.37	0.711	-.1982361	.2906795
state_27	-.0359834	.1325799	-0.27	0.786	-.2959154	.2239485
state_28	.0679452	.1186305	0.57	0.567	-.1646384	.3005287
state_29	-.0958816	.1363748	-0.70	0.482	-.3632535	.1714904
state_30	.0595816	.1266587	0.47	0.638	-.1887415	.3079047
state_31	-.0545955	.1305347	-0.42	0.676	-.3105178	.2013268
state_32	.0983601	.1188626	0.83	0.408	-.1346785	.3313986
state_33	.0703669	.1314686	0.54	0.593	-.1873864	.3281203
state_34	.0736822	.1215209	0.61	0.544	-.1645681	.3119325
state_35	.1333261	.1175253	1.13	0.257	-.0970906	.3637429
state_36	.056149	.1184076	0.47	0.635	-.1759974	.2882954
state_37	.0307212	.1215452	0.25	0.800	-.2075767	.269019
state_38	.1154202	.1210752	0.95	0.341	-.1219562	.3527965
state_39	.0626617	.1180676	0.53	0.596	-.1688183	.2941416
state_40	.0503717	.1324429	0.38	0.704	-.2092918	.3100351
state_41	.0704121	.1203691	0.58	0.559	-.16558	.3064042
state_42	-.2557383	.1564265	-1.63	0.102	-.5624227	.0509462
state_43	.0163814	.1201486	0.14	0.892	-.2191784	.2519413
state_44	.0728972	.1175644	0.62	0.535	-.157596	.3033904
state_45	.0376812	.124309	0.30	0.762	-.2060352	.2813976
state_46	.0665406	.1192939	0.56	0.577	-.1673435	.3004247
state_47	.0561053	.1381574	0.41	0.685	-.2147618	.3269723
state_48	.0943275	.1197309	0.79	0.431	-.1404133	.3290684
state_49	.0674351	.1207075	0.56	0.576	-.1692205	.3040907
state_50	.0973938	.1243632	0.78	0.434	-.1464289	.3412165
state_51	-.0464046	.1380825	-0.34	0.737	-.3171248	.2243156
polengage_scale	.1101144	.0178363	6.17	0.000	.0751452	.1450836
_cons	.6308111	.1190577	5.30	0.000	.3973901	.8642322

35 . mi estimate: regress dv_behavior c.auth_scale##c.polengage_scale i.sample_num

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 4,029
Average RVI = 0.0000
Largest FMI = 0.0000
Complete DF = 4022
DF adjustment: Small sample — DF: min = 4,020.00
avg = 4,020.00
max = 4,020.00
Model F test: Equal FMI — F(6, 4020.0) = 117.81
Within VCE type: OLS — Prob > F = 0.0000

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.0886194	.0367693	2.41	0.016	.0165311	.1607076
polengage_scale	.2431383	.030853	7.88	0.000	.1826493	.3036273
c.auth_scale#c.polengage_scale	-.1788169	.052023	-3.44	0.001	-.2808107	-.076823
sample_num						
3	.0305082	.0134575	2.27	0.023	.004124	.0568924
4	-.1468914	.012361	-11.88	0.000	-.1711257	-.122657
5	.0131838	.0153475	0.86	0.390	-.0169057	.0432733
_cons	.6706658	.0245997	27.26	0.000	.6224368	.7188949

36 . mi estimate: regress dv_behavior c.auth_scale##c.polengage_scale c.age i.male i.race c.education c.income i.sample_nu

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 4,029
Average RVI = 0.0013
Largest FMI = 0.0214
Complete DF = 3964
DF adjustment: **Small sample** — DF: min = 3,610.82
avg = 3,913.04
max = 3,960.60
Model F test: **Equal FMI** — F(64, 3962.0) = 15.80
Within VCE type: **OLS** — Prob > F = 0.0000

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.0704089	.0364198	1.93	0.053	-.0009943	.1418122
polengage_scale	.1943643	.0320106	6.07	0.000	.1316055	.2571232
c.auth_scale#c.polengage_scale	-.1621334	.0511717	-3.17	0.002	-.2624587	-.061808
age	.0446984	.0183024	2.44	0.015	.0088146	.0805823
1.male	-.0240082	.0078251	-3.07	0.002	-.03935	-.0086664
race						
1	-.0192203	.0148217	-1.30	0.195	-.0482791	.0098386
2	.0585929	.0183413	3.19	0.001	.0226335	.0945523
3	.031847	.0173525	1.84	0.067	-.0021737	.0658677
education	.0576995	.015845	3.64	0.000	.0266342	.0887647
income	.0666698	.0144385	4.62	0.000	.0383613	.0949783
sample_num						
3	.0108301	.0136125	0.80	0.426	-.0158581	.0375183
4	-.1557318	.0124403	-12.52	0.000	-.1801218	-.1313418
5	.0042465	.0156952	0.27	0.787	-.026525	.035018
state_1	.1101583	.1459435	0.75	0.450	-.1759737	.3962903
state_2	.0071499	.1204637	0.06	0.953	-.2290275	.2433274
state_3	.0984111	.1222118	0.81	0.421	-.1411937	.3380159
state_4	.0570497	.1187994	0.48	0.631	-.1758649	.2899642
state_5	.1316837	.1173078	1.12	0.262	-.0983066	.3616739
state_6	.0167802	.1199184	0.14	0.889	-.2183282	.2518885
state_7	.1198798	.1230418	0.97	0.330	-.1213522	.3611118
state_8	.1597008	.1282927	1.24	0.213	-.0918259	.4112275
state_9	.1182025	.1320862	0.89	0.371	-.1407615	.3771664
state_10	.0714162	.1174647	0.61	0.543	-.1588816	.301714
state_11	.075598	.1182545	0.64	0.523	-.1562482	.3074442
state_12	.2485938	.1287479	1.93	0.054	-.0038254	.501013
state_13	-.0150884	.1228372	-0.12	0.902	-.2559193	.2257424
state_14	.1436048	.1303007	1.10	0.270	-.1118587	.3990682

state_15	.0577669	.1183667	0.49	0.626	-.1742994	.2898331
state_16	.037057	.1196458	0.31	0.757	-.197517	.271631
state_17	.0197516	.1219691	0.16	0.871	-.2193774	.2588806
state_18	.0881239	.1205508	0.73	0.465	-.1482243	.3244721
state_19	.0459027	.1210827	0.38	0.705	-.1914884	.2832938
state_20	.1078351	.1198902	0.90	0.368	-.127218	.3428882
state_21	.08738	.1198011	0.73	0.466	-.1474985	.3222584
state_22	.1418441	.1304197	1.09	0.277	-.1138528	.3975409
state_23	.0662915	.1185951	0.56	0.576	-.1662227	.2988056
state_24	.0432455	.1210228	0.36	0.721	-.1940282	.2805193
state_25	.0326463	.1194952	0.27	0.785	-.2016325	.266925
state_26	.0449969	.1245355	0.36	0.718	-.1991637	.2891574
state_27	-.0323556	.1324264	-0.24	0.807	-.2919865	.2272753
state_28	.0669774	.1184865	0.57	0.572	-.1653238	.2992786
state_29	-.0997585	.1362167	-0.73	0.464	-.3668206	.1673037
state_30	.0595325	.1265058	0.47	0.638	-.1884908	.3075558
state_31	-.0513786	.1303821	-0.39	0.694	-.3070018	.2042445
state_32	.0974818	.1187183	0.82	0.412	-.1352738	.3302373
state_33	.0657729	.1313179	0.50	0.616	-.1916849	.3232307
state_34	.0728349	.1213737	0.60	0.548	-.1651267	.3107965
state_35	.1323084	.1173825	1.13	0.260	-.0978282	.362445
state_36	.0556249	.1182636	0.47	0.638	-.1762393	.2874891
state_37	.0280602	.1214007	0.23	0.817	-.2099545	.2660748
state_38	.1121136	.1209326	0.93	0.354	-.1249832	.3492105
state_39	.0611098	.1179249	0.52	0.604	-.1700902	.2923098
state_40	.0469961	.1322876	0.36	0.722	-.2123628	.3063551
state_41	.0711293	.1202234	0.59	0.554	-.164577	.3068357
state_42	-.2486015	.1562579	-1.59	0.112	-.5549555	.0577525
state_43	.0144983	.1200039	0.12	0.904	-.2207778	.2497743
state_44	.0723659	.1174211	0.62	0.538	-.1578465	.3025782
state_45	.0366759	.1241587	0.30	0.768	-.2067459	.2800977
state_46	.0651653	.1191501	0.55	0.584	-.1684368	.2987673
state_47	.053255	.1379947	0.39	0.700	-.2172929	.3238029
state_48	.0947703	.119585	0.79	0.428	-.1396845	.3292252
state_49	.0649735	.1205632	0.54	0.590	-.1713991	.3013461
state_50	.0968613	.1242143	0.78	0.436	-.1466693	.3403919
state_51	-.0448718	.1379183	-0.33	0.745	-.3152699	.2255263
_cons	.576003	.1201704	4.79	0.000	.3404006	.8116054

37 . mi estimate: regress dv_behavior c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.
> n##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 4,029
Average RVI = 0.0040
Largest FMI = 0.0700
Complete DF = 3957
DF adjustment: **Small sample** — DF: min = 2,111.97
avg = 3,818.01
max = 3,950.25
Model F test: **Equal FMI** — F(71, 3954.8) = 15.08
Within VCE type: **OLS** — Prob > F = 0.0000

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.0367659	.0370423	0.99	0.321	-.035858	.1093898
polengage_scale	.4026752	.0718669	5.60	0.000	.2617753	.5435751
c.auth_scale#c.polengage_scale	-.1207176	.0520662	-2.32	0.020	-.2227969	-.0186383
age	.3338402	.0517109	6.46	0.000	.2324427	.4352376
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.4426272	.0732594	-6.04	0.000	-.5862792	-.2989752

1.male	.0066576	.0220199	0.30	0.762	-.0365173	.0498326
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.0470879	.0317922	-1.48	0.139	-.1094226	.0152467
race						
1	-.034041	.0401753	-0.85	0.397	-.1128073	.0447253
2	.0879404	.0479182	1.84	0.067	-.0060063	.1818871
3	.1030468	.0457485	2.25	0.024	.0133536	.19274
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.0248327	.0603273	0.41	0.681	-.0934432	.1431086
2	-.040581	.0744247	-0.55	0.586	-.1864955	.1053335
3	-.1222166	.0708557	-1.72	0.085	-.261134	.0167008
education	.0729715	.043456	1.68	0.093	-.0122295	.1581725
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	-.0297745	.0625925	-0.48	0.634	-.1524944	.0929453
income	.1033426	.0411838	2.51	0.012	.0225775	.1841077
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0599814	.0581055	-1.03	0.302	-.1739221	.0539593
sample_num						
3	.0122216	.0135802	0.90	0.368	-.0144032	.0388464
4	-.1527008	.0124123	-12.30	0.000	-.177036	-.1283656
5	-.0002833	.0156648	-0.02	0.986	-.0309952	.0304285
state_1	.1200439	.1451191	0.83	0.408	-.1644717	.4045596
state_2	.0116089	.1197683	0.10	0.923	-.2232051	.2464228
state_3	.1019174	.1215034	0.84	0.402	-.1362983	.3401332
state_4	.0629931	.1181404	0.53	0.594	-.1686294	.2946155
state_5	.1362771	.1166251	1.17	0.243	-.0923745	.3649287
state_6	.0200526	.1192456	0.17	0.866	-.2137365	.2538417
state_7	.1256231	.1223928	1.03	0.305	-.1143364	.3655826
state_8	.164833	.1275245	1.29	0.196	-.0851874	.4148535
state_9	.1214752	.1314	0.92	0.355	-.1361433	.3790938
state_10	.0779239	.1167995	0.67	0.505	-.1510695	.3069173
state_11	.08004	.1176083	0.68	0.496	-.1505391	.3106191
state_12	.2583585	.128042	2.02	0.044	.0073234	.5093936
state_13	-.0125625	.1221359	-0.10	0.918	-.2520184	.2268933
state_14	.1450564	.1295566	1.12	0.263	-.1089481	.3990608
state_15	.0613872	.1176885	0.52	0.602	-.1693492	.2921236
state_16	.0349227	.1189422	0.29	0.769	-.1982716	.268117
state_17	.0229152	.12127	0.19	0.850	-.214843	.2606734
state_18	.0962122	.1198931	0.80	0.422	-.1388465	.3312709
state_19	.0540787	.1204065	0.45	0.653	-.1819864	.2901438
state_20	.108403	.1191988	0.91	0.363	-.1252945	.3421004
state_21	.0924213	.1190982	0.78	0.438	-.141079	.3259215
state_22	.1388921	.1297044	1.07	0.284	-.1154022	.3931865
state_23	.0697839	.11792	0.59	0.554	-.1614063	.3009742
state_24	.0575931	.1203137	0.48	0.632	-.1782901	.2934762
state_25	.033491	.1188077	0.28	0.778	-.1994398	.2664217
state_26	.0495724	.1238265	0.40	0.689	-.193198	.2923428
state_27	-.021946	.1317358	-0.17	0.868	-.2802232	.2363311
state_28	.0729592	.1178082	0.62	0.536	-.1580119	.3039302
state_29	-.0877901	.1354494	-0.65	0.517	-.3533478	.1777676
state_30	.0622156	.1257897	0.49	0.621	-.1844036	.3088347
state_31	-.0571627	.1296529	-0.44	0.659	-.3113559	.1970306
state_32	.1078641	.1180417	0.91	0.361	-.1235647	.3392929
state_33	.0780407	.1308409	0.60	0.551	-.1784825	.334564
state_34	.0761606	.1207182	0.63	0.528	-.1605156	.3128369

state_35	.1339861	.1167162	1.15	0.251	-.0948439	.3628162
state_36	.0635035	.1175919	0.54	0.589	-.1670435	.2940505
state_37	.0374509	.1207098	0.31	0.756	-.1992089	.2741108
state_38	.1151749	.1202452	0.96	0.338	-.1205741	.3509238
state_39	.0677143	.1172479	0.58	0.564	-.1621582	.2975868
state_40	.0537204	.1315207	0.41	0.683	-.2041347	.3115755
state_41	.0769336	.119522	0.64	0.520	-.1573976	.3112647
state_42	-.245201	.1555379	-1.58	0.115	-.5501433	.0597413
state_43	.0195644	.1193358	0.16	0.870	-.2144017	.2535304
state_44	.0758343	.116749	0.65	0.516	-.1530601	.3047287
state_45	.0392792	.1234483	0.32	0.750	-.2027495	.281308
state_46	.0737052	.1184957	0.62	0.534	-.1586136	.306024
state_47	.0535106	.1372468	0.39	0.697	-.215571	.3225923
state_48	.0961175	.1189052	0.81	0.419	-.1370043	.3292394
state_49	.0723407	.1198802	0.60	0.546	-.1626926	.307374
state_50	.0926873	.1235224	0.75	0.453	-.1494868	.3348615
state_51	-.0291579	.1371451	-0.21	0.832	-.2980401	.2397243
_cons	.45089	.1260296	3.58	0.000	.2038001	.6979798

```
38 .
39 . *Restrictions, Authoritarianism
40 . mi estimate: regress dv_restrictions c.auth_scale i.sample_num
```

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 6,870
Average RVI = 0.0000
Largest FMI = 0.0000
Complete DF = 6866
DF adjustment: **Small sample** — DF: min = 6,864.00
avg = 6,864.00
max = 6,864.00
Model F test: **Equal FMI** — F(3, 6864.0) = 113.78
Within VCE type: **OLS** — Prob > F = 0.0000

dv_restric~s	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.1548316	.0108907	-14.22	0.000	-.1761807	-.1334825
sample_num						
4	-.0992496	.0114058	-8.70	0.000	-.1216085	-.0768906
6	-.0095801	.0106007	-0.90	0.366	-.0303607	.0112004
_cons	.7247066	.0112775	64.26	0.000	.7025991	.7468141

```
41 . mi estimate: regress dv_restrictions c.auth_scale c.age i.male i.race c.education c.income i.sample_num state_1-state_
```

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 6,870
Average RVI = 0.0022
Largest FMI = 0.0917
Complete DF = 6808
DF adjustment: **Small sample** — DF: min = 1,972.00
avg = 6,703.06
max = 6,805.80
Model F test: **Equal FMI** — F(61, 6805.8) = 13.74
Within VCE type: **OLS** — Prob > F = 0.0000

dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.1523805	.0111173	-13.71	0.000	-.1741738	-.1305872
age	-.0812396	.018438	-4.41	0.000	-.1173842	-.0450951
1.male	-.0465439	.0074766	-6.23	0.000	-.0612005	-.0318873
race						
1	-.0206333	.0142772	-1.45	0.148	-.0486211	.0073545
2	.1248806	.0179639	6.95	0.000	.0896657	.1600955
3	.0327445	.0178329	1.84	0.066	-.0022136	.0677026
education	.0730282	.0143009	5.11	0.000	.0449934	.1010629
income	.0325874	.015733	2.07	0.038	.0017323	.0634426
sample_num						
4	-.0881446	.0112629	-7.83	0.000	-.1102235	-.0660657
6	.0219285	.010765	2.04	0.042	.0008256	.0430313
state_1	.1235184	.0967332	1.28	0.202	-.066109	.3131459
state_2	.0211428	.0801691	0.26	0.792	-.1360137	.1782993
state_3	.0865614	.0815141	1.06	0.288	-.0732317	.2463546
state_4	.0204963	.0774313	0.26	0.791	-.1312933	.1722858
state_5	.1367982	.0754843	1.81	0.070	-.0111746	.2847711
state_6	.0766391	.0786805	0.97	0.330	-.0775993	.2308775
state_7	.0991329	.083126	1.19	0.233	-.0638201	.2620859
state_8	.284214	.096059	2.96	0.003	.0959083	.4725197
state_9	.0965637	.0943131	1.02	0.306	-.0883195	.2814469
state_10	.0686961	.0755979	0.91	0.364	-.0794994	.2168916
state_11	.0591726	.0770634	0.77	0.443	-.0918958	.210241
state_12	.1643034	.0907712	1.81	0.070	-.0136365	.3422433
state_13	.1090525	.0818518	1.33	0.183	-.0514027	.2695076
state_14	.0902122	.0861061	1.05	0.295	-.0785827	.259007
state_15	.0784376	.0769211	1.02	0.308	-.0723519	.2292271
state_16	.0547942	.0785506	0.70	0.485	-.0991895	.2087779
state_17	.1177095	.0821726	1.43	0.152	-.0433745	.2787935
state_18	.1098333	.080647	1.36	0.173	-.04826	.2679267
state_19	-.0460658	.0818767	-0.56	0.574	-.2065697	.1144382
state_20	.1219673	.078494	1.55	0.120	-.0319055	.2758401
state_21	.1405253	.0786708	1.79	0.074	-.0136941	.2947448
state_22	.1304382	.0877066	1.49	0.137	-.041494	.3023705
state_23	.0715666	.0770125	0.93	0.353	-.079402	.2225351
state_24	.080752	.0799515	1.01	0.313	-.075978	.237482
state_25	.0153065	.0781904	0.20	0.845	-.1379711	.1685841
state_26	-.0264428	.0859314	-0.31	0.758	-.1948952	.1420096
state_27	-.0550561	.091334	-0.60	0.547	-.2340994	.1239871
state_28	.0590809	.0774115	0.76	0.445	-.0926699	.2108317
state_29	-.0311746	.1072506	-0.29	0.771	-.2414194	.1790701
state_30	.0642892	.086084	0.75	0.455	-.1044623	.2330408
state_31	.0065873	.0891441	0.07	0.941	-.1681631	.1813377
state_32	.060616	.0775022	0.78	0.434	-.0913126	.2125446
state_33	-.0591922	.0867194	-0.68	0.495	-.2291892	.1108049
state_34	.0304072	.0818864	0.37	0.710	-.1301157	.1909302
state_35	.1655991	.0758331	2.18	0.029	.0169425	.3142557
state_36	.0850066	.0768677	1.11	0.269	-.0656782	.2356914
state_37	.0811346	.0832043	0.98	0.330	-.0819718	.2442411
state_38	.0755668	.0803517	0.94	0.347	-.0819476	.2330813
state_39	.0510653	.0760511	0.67	0.502	-.0980186	.2001493
state_40	.1959593	.0958289	2.04	0.041	.0081048	.3838139
state_41	.0991831	.0801603	1.24	0.216	-.057956	.2563223
state_42	-.0713762	.0950185	-0.75	0.453	-.2576421	.1148898
state_43	.0623861	.0787013	0.79	0.428	-.091893	.2166652
state_44	.0646239	.075772	0.85	0.394	-.083913	.2131608
state_45	.0079954	.0833841	0.10	0.924	-.1554634	.1714543
state_46	.0591875	.0778565	0.76	0.447	-.0934356	.2118106
state_47	.1062752	.0967387	1.10	0.272	-.0833629	.2959134
state_48	.0573538	.0778472	0.74	0.461	-.095251	.2099587

state_49	.0551491	.0782876	0.70	0.481	-.0983192	.2086173
state_50	.1097716	.0847879	1.29	0.195	-.0564392	.2759825
state_51	0	(omitted)				
polengage_scale	-.0243603	.0164563	-1.48	0.139	-.0566199	.0078993
_cons	.6438196	.0774032	8.32	0.000	.4920852	.795554

42 . mi estimate: regress dv_restrictions c.auth_scale##c.polengage_scale i.sample_num

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 6,870
Average RVI = 0.0000
Largest FMI = 0.0000
Complete DF = 6864
DF adjustment: Small sample — DF: min = 6,862.00
avg = 6,862.00
max = 6,862.00
Model F test: Equal FMI — F(5, 6862.0) = 96.97
Within VCE type: OLS — Prob > F = 0.0000

dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.208611	.0342924	6.08	0.000	.1413873	.2758348
polengage_scale	.2225571	.0278265	8.00	0.000	.1680085	.2771058
c.auth_scale#c.polengage_scale	-.4942001	.0435417	-11.35	0.000	-.5795552	-.4088449
sample_num						
4	-.0965049	.0113864	-8.48	0.000	-.1188257	-.074184
6	-.0115115	.0106982	-1.08	0.282	-.0324834	.0094604
_cons	.5532566	.0238351	23.21	0.000	.5065325	.5999807

43 . mi estimate: regress dv_restrictions c.auth_scale##c.polengage_scale c.age i.male i.race c.education c.income i.sample

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 6,870
Average RVI = 0.0022
Largest FMI = 0.0876
Complete DF = 6807
DF adjustment: Small sample — DF: min = 2,101.01
avg = 6,704.76
max = 6,804.77
Model F test: Equal FMI — F(62, 6804.8) = 15.74
Within VCE type: OLS — Prob > F = 0.0000

dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.1998433	.0336468	5.94	0.000	.133885	.2658016
polengage_scale	.2317673	.0282916	8.19	0.000	.1763069	.2872277
c.auth_scale#c.polengage_scale	-.4722484	.0426234	-11.08	0.000	-.5558035	-.3886933
age	-.0965246	.0183342	-5.26	0.000	-.1324656	-.0605835
1.male	-.0454781	.0074178	-6.13	0.000	-.0600195	-.0309366
race						
1	-.0204592	.0141514	-1.45	0.148	-.0482002	.0072819
2	.1206115	.0178095	6.77	0.000	.0856994	.1555236
3	.03429	.0176763	1.94	0.052	-.0003611	.068941
education	.0675639	.0141776	4.77	0.000	.0397709	.0953569
income	.0344796	.0155602	2.22	0.027	.0039646	.0649947

sample_num						
4	-.0821036	.0111769	-7.35	0.000	-.1040138	-.0601933
6	.0155105	.0106852	1.45	0.147	-.0054358	.0364568
state_1	.1252098	.0958796	1.31	0.192	-.0627442	.3131638
state_2	.0210276	.079462	0.26	0.791	-.1347428	.176798
state_3	.0838122	.080795	1.04	0.300	-.0745713	.2421957
state_4	.0236356	.0767485	0.31	0.758	-.1268154	.1740867
state_5	.1340436	.0748185	1.79	0.073	-.0126241	.2807113
state_6	.0742262	.0779866	0.95	0.341	-.078652	.2271044
state_7	.0928013	.0823945	1.13	0.260	-.0687176	.2543202
state_8	.2853939	.0952109	3.00	0.003	.0987508	.472037
state_9	.1009522	.0934816	1.08	0.280	-.082301	.2842054
state_10	.0676887	.0749309	0.90	0.366	-.0791992	.2145767
state_11	.0595157	.0763834	0.78	0.436	-.0902196	.209251
state_12	.1652397	.08997	1.84	0.066	-.0111297	.3416091
state_13	.1011879	.0811327	1.25	0.212	-.0578576	.2602334
state_14	.0856871	.0853472	1.00	0.315	-.0816202	.2529944
state_15	.0792101	.0762424	1.04	0.299	-.0702489	.228669
state_16	.051688	.077858	0.66	0.507	-.1009381	.2043142
state_17	.1104593	.0814505	1.36	0.175	-.0492092	.2701277
state_18	.109617	.0799355	1.37	0.170	-.0470816	.2663156
state_19	-.0446505	.0811543	-0.55	0.582	-.2037382	.1144372
state_20	.1173377	.0778026	1.51	0.132	-.0351797	.2698551
state_21	.1383475	.077977	1.77	0.076	-.0145118	.2912069
state_22	.1251987	.0869339	1.44	0.150	-.0452189	.2956163
state_23	.0689171	.0763333	0.90	0.367	-.0807199	.2185542
state_24	.083023	.0792464	1.05	0.295	-.0723246	.2383707
state_25	.0123697	.0775009	0.16	0.873	-.1395563	.1642956
state_26	-.0349999	.0851772	-0.41	0.681	-.2019739	.131974
state_27	-.0419376	.0905364	-0.46	0.643	-.2194173	.135542
state_28	.0602794	.0767285	0.79	0.432	-.0901325	.2106912
state_29	-.0391962	.1063065	-0.37	0.712	-.2475902	.1691978
state_30	.0635241	.0853252	0.74	0.457	-.1037399	.2307881
state_31	.0072268	.0883575	0.08	0.935	-.1659815	.1804352
state_32	.0585128	.0768187	0.76	0.446	-.0920759	.2091015
state_33	-.0611325	.0859562	-0.71	0.477	-.2296335	.1073686
state_34	.0376635	.0811665	0.46	0.643	-.1214482	.1967752
state_35	.1625098	.0751644	2.16	0.031	.0151641	.3098554
state_36	.0846176	.0761893	1.11	0.267	-.0647373	.2339725
state_37	.0769364	.0824715	0.93	0.351	-.0847336	.2386064
state_38	.0711358	.0796437	0.89	0.372	-.0849908	.2272623
state_39	.0510954	.0753801	0.68	0.498	-.0966732	.198864
state_40	.1976905	.0949835	2.08	0.037	.0114931	.3838878
state_41	.0993822	.0794535	1.25	0.211	-.0563714	.2551358
state_42	-.0615008	.0941843	-0.65	0.514	-.2461314	.1231299
state_43	.0590338	.0780074	0.76	0.449	-.0938851	.2119527
state_44	.0657517	.0751036	0.88	0.381	-.0814748	.2129783
state_45	.0093838	.0826483	0.11	0.910	-.1526326	.1714003
state_46	.0580523	.0771699	0.75	0.452	-.0932248	.2093294
state_47	.1027083	.0958862	1.07	0.284	-.0852587	.2906753
state_48	.0585927	.0771604	0.76	0.448	-.0926659	.2098512
state_49	.0549303	.0775969	0.71	0.479	-.0971839	.2070444
state_50	.102747	.084042	1.22	0.222	-.0620017	.2674956
state_51	0	(omitted)				
_cons	.4573272	.0785438	5.82	0.000	.3033569	.6112976

```
44 . mi estimate: regress dv_restrictions c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scal
> ation##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      6,870
                                                Average RVI       =     0.0044
                                                Largest FMI       =     0.0853
                                                Complete DF       =       6800
DF adjustment:   Small sample                   DF:     min       =   2,174.22
                                                        avg       =   6,549.62
                                                        max       =   6,797.96
Model F test:       Equal FMI                   F(  69, 6797.4)   =      15.39
Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.1982025	.0342422	5.79	0.000	.1310771	.2653279
polengage_scale	.4185124	.0669678	6.25	0.000	.287234	.5497908
c.auth_scale#c.polengage_scale	-.4771861	.04359	-10.95	0.000	-.5626362	-.3917361
age	.1032363	.0524726	1.97	0.049	.0003607	.2061119
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.2919973	.0716903	-4.07	0.000	-.4325495	-.1514451
1.male	.0668953	.0223694	2.99	0.003	.0230438	.1107469
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	-.1495827	.0288798	-5.18	0.000	-.2061968	-.0929685
race						
1	-.01038	.0388059	-0.27	0.789	-.0864519	.0656918
2	.006421	.0468799	0.14	0.891	-.0854785	.0983204
3	.0387747	.0466565	0.83	0.406	-.0526866	.130236
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.012385	.0524737	-0.24	0.813	-.1152499	.0904799
2	.1828785	.0657748	2.78	0.005	.0539393	.3118177
3	-.004319	.0658323	-0.07	0.948	-.133371	.1247329
education	.038196	.0430641	0.89	0.375	-.0462251	.1226172
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.0417559	.0554505	0.75	0.451	-.0669472	.1504589
income	.099619	.0453578	2.20	0.028	.0106765	.1885615
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0947405	.0602526	-1.57	0.116	-.2128992	.0234182
sample_num						
4	-.0797173	.0111313	-7.16	0.000	-.1015382	-.0578964
6	.0142547	.0107762	1.32	0.186	-.0068699	.0353794
state_1	.1240173	.0953982	1.30	0.194	-.062993	.3110275
state_2	.0233894	.0790689	0.30	0.767	-.1316105	.1783892
state_3	.0821177	.0803877	1.02	0.307	-.0754673	.2397028
state_4	.0225871	.0763724	0.30	0.767	-.1271267	.1723008
state_5	.1315333	.0744544	1.77	0.077	-.0144207	.2774872
state_6	.0733953	.0776101	0.95	0.344	-.0787448	.2255353
state_7	.0930679	.0819915	1.14	0.256	-.067661	.2537969
state_8	.2732221	.0947601	2.88	0.004	.0874626	.4589816

state_9	.1027545	.0930954	1.10	0.270	-.0797415	.2852506
state_10	.0673181	.0745718	0.90	0.367	-.0788661	.2135022
state_11	.0637246	.0760211	0.84	0.402	-.0853005	.2127498
state_12	.1625155	.0895745	1.81	0.070	-.0130785	.3381096
state_13	.100022	.0807433	1.24	0.215	-.0582601	.2583042
state_14	.0863937	.084935	1.02	0.309	-.0801056	.2528929
state_15	.082832	.075881	1.09	0.275	-.0659185	.2315824
state_16	.0481209	.0774846	0.62	0.535	-.103773	.2000149
state_17	.1072039	.0810532	1.32	0.186	-.0516857	.2660934
state_18	.1145232	.0795437	1.44	0.150	-.0414073	.2704538
state_19	-.0369432	.0807556	-0.46	0.647	-.1952495	.1213631
state_20	.1141451	.0774196	1.47	0.140	-.0376216	.2659118
state_21	.1363348	.0775901	1.76	0.079	-.015766	.2884357
state_22	.1258207	.0865286	1.45	0.146	-.0438024	.2954438
state_23	.0660586	.0759568	0.87	0.385	-.0828406	.2149578
state_24	.0914904	.0788496	1.16	0.246	-.0630794	.2460602
state_25	.0075235	.0771184	0.10	0.922	-.1436528	.1586998
state_26	-.0303459	.0847785	-0.36	0.720	-.1965383	.1358465
state_27	-.0381998	.0900995	-0.42	0.672	-.2148229	.1384234
state_28	.0610487	.076351	0.80	0.424	-.0886231	.2107205
state_29	-.0403775	.1057781	-0.38	0.703	-.2477357	.1669806
state_30	.0610406	.0848956	0.72	0.472	-.1053813	.2274625
state_31	.0071045	.0879304	0.08	0.936	-.1652667	.1794757
state_32	.0601869	.076455	0.79	0.431	-.0896888	.2100625
state_33	-.0577224	.0855445	-0.67	0.500	-.2254163	.1099715
state_34	.0437128	.0807833	0.54	0.588	-.1146478	.2020733
state_35	.1605565	.0748	2.15	0.032	.013925	.307188
state_36	.0879078	.0758124	1.16	0.246	-.0607082	.2365237
state_37	.078821	.0820663	0.96	0.337	-.0820545	.2396966
state_38	.0659131	.079259	0.83	0.406	-.0894593	.2212856
state_39	.0494361	.0750118	0.66	0.510	-.0976105	.1964828
state_40	.1951315	.094535	2.06	0.039	.0098134	.3804497
state_41	.0930807	.0790515	1.18	0.239	-.061885	.2480465
state_42	-.0597531	.0937336	-0.64	0.524	-.2435002	.123994
state_43	.0615299	.0776265	0.79	0.428	-.0906424	.2137022
state_44	.0645122	.0747308	0.86	0.388	-.0819836	.211008
state_45	.0132268	.0822607	0.16	0.872	-.1480299	.1744834
state_46	.0602913	.0767847	0.79	0.432	-.0902308	.2108134
state_47	.1000887	.0954507	1.05	0.294	-.0870244	.2872019
state_48	.0584947	.0767836	0.76	0.446	-.0920251	.2090145
state_49	.0584114	.0772236	0.76	0.449	-.0929709	.2097938
state_50	.1033619	.0836308	1.24	0.217	-.0605806	.2673044
state_51	0	(omitted)				
_cons	.3366418	.0880965	3.82	0.000	.163945	.5093385

```
45 .
46 . *Concern, Latent NSC
47 . mi estimate: regress dv_concern c.latentnsc2 i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        966
                                                Average RVI       =     0.0000
                                                Largest FMI       =     0.0000
                                                Complete DF       =        963
DF adjustment:   Small sample                   DF:     min       =     961.01
                                                        avg       =     961.01
                                                        max       =     961.01
Model F test:       Equal FMI                   F(   2,  961.0)   =       3.68
Within VCE type:          OLS                   Prob > F          =     0.0256
```

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.1295556	.0548211	-2.36	0.018	-.2371386	-.0219727
5.sample_num	-.0367518	.0223862	-1.64	0.101	-.0806834	.0071797
_cons	.5788026	.0310225	18.66	0.000	.517923	.6396823

48 . mi estimate: regress dv_concern c.latentnsc2 c.age i.male i.race c.education c.income i.sample_num polengage_scale

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 965
Average RVI = 0.0071
Largest FMI = 0.0237
Complete DF = 954
DF adjustment: Small sample — DF: min = 910.02
avg = 941.77
max = 951.54
Model F test: Equal FMI — F(10, 951.7) = 3.65
Within VCE type: OLS — Prob > F = 0.0001

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.1165662	.0573018	-2.03	0.042	-.2290187	-.0041136
age	.0138878	.0546539	0.25	0.799	-.0933683	.121144
1.male	-.05045	.0228317	-2.21	0.027	-.0952563	-.0056437
race						
1	-.0450214	.0478093	-0.94	0.347	-.1388469	.048804
2	.0956522	.0595398	1.61	0.109	-.021199	.2125034
3	.0301881	.0579341	0.52	0.602	-.0835086	.1438847
education	.1263697	.0507914	2.49	0.013	.0266927	.2260468
income	-.0881743	.0428639	-2.06	0.040	-.1722947	-.004054
5.sample_num	-.0231249	.0263392	-0.88	0.380	-.0748147	.0285649
polengage_scale	.09337	.0608352	1.53	0.125	-.0260169	.2127569
_cons	.4947325	.0692623	7.14	0.000	.3588069	.6306581

49 . mi estimate: regress dv_concern c.latentnsc2##c.polengage_scale i.sample_num

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 965
Average RVI = 0.0000
Largest FMI = 0.0000
Complete DF = 960
DF adjustment: Small sample — DF: min = 958.01
avg = 958.01
max = 958.01
Model F test: Equal FMI — F(4, 958.0) = 3.73
Within VCE type: OLS — Prob > F = 0.0051

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.4207813	.2322453	1.81	0.070	-.0349869	.8765495
polengage_scale	.4129302	.1512812	2.73	0.006	.1160494	.709811
c.latentnsc2#c.polengage_scale	-.7106031	.2971453	-2.39	0.017	-1.293734	-.1274724
5.sample_num	-.0413563	.0226236	-1.83	0.068	-.0857538	.0030411
_cons	.2613927	.1202521	2.17	0.030	.0254049	.4973805

```
50 . mi estimate: regress dv_concern c.latentnsc2##c.polengage_scale c.age i.male i.race c.education c.income i.sample_num

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =        965
                                              Average RVI       =     0.0067
                                              Largest FMI       =     0.0242
                                              Complete DF       =        953
DF adjustment:   Small sample                 DF:     min       =     907.70
                                                      avg       =     941.78
                                                      max       =     950.50
Model F test:       Equal FMI                 F(  11,  950.8)   =       3.85
Within VCE type:          OLS                 Prob > F          =     0.0000
```

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.4208158	.2320095	1.81	0.070	-.0344943	.8761258
polengage_scale	.4310728	.1537751	2.80	0.005	.1292948	.7328508
c.latentnsc2#c.polengage_scale	-.7063315	.2955364	-2.39	0.017	-1.286311	-.1263522
age	.0110475	.0545319	0.20	0.840	-.0959694	.1180645
1.male	-.0511353	.0227774	-2.25	0.025	-.095835	-.0064355
race						
1	-.05257	.0478103	-1.10	0.272	-.1463976	.0412576
2	.0868909	.0595259	1.46	0.145	-.0299336	.2037153
3	.026572	.0578426	0.46	0.646	-.0869456	.1400895
education	.1208721	.0507149	2.38	0.017	.021345	.2203992
income	-.0853745	.0427656	-2.00	0.046	-.1693019	-.001447
5.sample_num	-.0236414	.0262751	-0.90	0.368	-.0752054	.0279227
_cons	.2444755	.1254188	1.95	0.052	-.0016547	.4906057

```
51 . mi estimate: regress dv_concern c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.r
   > ##c.polengage_scale c.income##c.polengage_scale i.sample_num

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =        965
                                              Average RVI       =     0.0112
                                              Largest FMI       =     0.0415
                                              Complete DF       =        946
DF adjustment:   Small sample                 DF:     min       =     850.29
                                                      avg       =     923.19
                                                      max       =     942.89
Model F test:       Equal FMI                 F(  18,  943.7)   =       2.70
Within VCE type:          OLS                 Prob > F          =     0.0002
```

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.3410344	.2394253	1.42	0.155	-.1288337	.8109026
polengage_scale	.2721278	.2842061	0.96	0.339	-.2856298	.8298853
c.latentnsc2#c.polengage_scale	-.5889081	.3060059	-1.92	0.055	-1.189439	.0116233
age	.1872601	.2037141	0.92	0.358	-.2125258	.5870459
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.2445519	.2655462	-0.92	0.357	-.7656824	.2765785
1.male	-.1650795	.0849524	-1.94	0.052	-.3317974	.0016384
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.1564718	.112463	1.39	0.164	-.0642352	.3771788

race						
1	-.1355216	.158605	-0.85	0.393	-.4467892	.175746
2	.2181534	.2067544	1.06	0.292	-.1876464	.6239531
3	.1352213	.1889212	0.72	0.474	-.2355495	.505992
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	.1144353	.2234904	0.51	0.609	-.3241758	.5530465
2	-.1953255	.297129	-0.66	0.511	-.7785178	.3878667
3	-.1645444	.2690128	-0.61	0.541	-.6925088	.36342
education	-.011299	.1792973	-0.06	0.950	-.3631737	.3405757
polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.1822275	.2418836	0.75	0.451	-.2924722	.6569271
income	-.055832	.1448773	-0.39	0.700	-.3401601	.2284961
polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0446706	.1896675	-0.24	0.814	-.4168985	.3275573
5.sample_num	-.0203423	.0263493	-0.77	0.440	-.0720525	.0313679
_cons	.3566723	.2037249	1.75	0.080	-.0431391	.7564836

```
52 .
53 . *Behaviors, Latent NSC
54 . mi estimate: regress dv_behavior c.latentnsc2 i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        971
                                                Average RVI       =     0.0000
                                                Largest FMI       =     0.0000
                                                Complete DF       =        968
DF adjustment:   Small sample                   DF:     min       =     966.01
                                                        avg       =     966.01
                                                        max       =     966.01
Model F test:       Equal FMI                   F(   2,  966.0)   =      10.50
Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.1434678	.0341745	-4.20	0.000	-.2105326	-.0764031
5.sample_num	.0182949	.0139248	1.31	0.189	-.0090314	.0456211
_cons	.8912888	.0192576	46.28	0.000	.8534972	.9290803

```
55 . mi estimate: regress dv_behavior c.latentnsc2 c.age i.male i.race c.education c.income i.sample_num polengage_scale

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        970
                                                Average RVI       =     0.0095
                                                Largest FMI       =     0.0201
                                                Complete DF       =        959
DF adjustment:   Small sample                   DF:     min       =     923.30
                                                        avg       =     942.68
                                                        max       =     956.18
Model F test:       Equal FMI                   F(  10,  956.5)   =       6.93
Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.1094668	.0351772	-3.11	0.002	-.1785002	-.0404334
age	-.0348343	.0334073	-1.04	0.297	-.1003951	.0307265
1.male	-.0587693	.0141542	-4.15	0.000	-.0865465	-.0309921
race						
1	-.0484991	.0292692	-1.66	0.098	-.1059405	.0089422
2	-.010748	.0361734	-0.30	0.766	-.0817382	.0602422
3	-.0277289	.0359329	-0.77	0.440	-.0982486	.0427908
education	.0579775	.0320662	1.81	0.071	-.0049519	.1209069
income	.0882629	.0269353	3.28	0.001	.0354021	.1411237
5.sample_num	-.0230041	.016314	-1.41	0.159	-.0550197	.0090115
polengage_scale	.1149322	.0375056	3.06	0.002	.0413294	.1885351
_cons	.8231528	.042848	19.21	0.000	.7390648	.9072408

56 . mi estimate: regress dv_behavior c.latentnsc2##c.polengage_scale i.sample_num

Multiple-imputation estimates	Imputations	=	25
Linear regression	Number of obs	=	970
	Average RVI	=	0.0000
	Largest FMI	=	0.0000
	Complete DF	=	965
DF adjustment: Small sample	DF: min	=	963.01
	avg	=	963.01
	max	=	963.01
Model F test: Equal FMI	F(4, 963.0)	=	7.56
Within VCE type: OLS	Prob > F	=	0.0000

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.0301302	.1428296	0.21	0.833	-.2501629	.3104233
polengage_scale	.1965101	.0931053	2.11	0.035	.0137973	.3792228
c.latentnsc2#c.polengage_scale	-.2133615	.1838774	-1.16	0.246	-.5742082	.1474851
5.sample_num	.0132093	.0140194	0.94	0.346	-.0143029	.0407214
_cons	.742237	.0735631	10.09	0.000	.5978746	.8865995

57 . mi estimate: regress dv_behavior c.latentnsc2##c.polengage_scale c.age i.male i.race c.education c.income i.sample_nur

Multiple-imputation estimates	Imputations	=	25
Linear regression	Number of obs	=	970
	Average RVI	=	0.0087
	Largest FMI	=	0.0202
	Complete DF	=	958
DF adjustment: Small sample	DF: min	=	922.31
	avg	=	943.47
	max	=	955.72
Model F test: Equal FMI	F(11, 955.7)	=	6.48
Within VCE type: OLS	Prob > F	=	0.0000

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.0756825	.1417568	0.53	0.594	-.202508	.3538731
polengage_scale	.2312294	.0940595	2.46	0.014	.0466422	.4158166
c.latentnsc2#c.polengage_scale	-.2449043	.1816448	-1.35	0.178	-.601373	.1115644
age	-.0363318	.0334107	-1.09	0.277	-.1018995	.0292359
1.male	-.059273	.0141527	-4.19	0.000	-.0870473	-.0314987
race						
1	-.0505687	.0292956	-1.73	0.085	-.1080618	.0069244
2	-.0137591	.0362326	-0.38	0.704	-.0848656	.0573474
3	-.0283749	.0359214	-0.79	0.430	-.098872	.0421222
education	.056445	.0320757	1.76	0.079	-.0065032	.1193932
income	.0886068	.0269314	3.29	0.001	.0357535	.1414602
5.sample_num	-.0231023	.0163069	-1.42	0.157	-.055104	.0088993
_cons	.7374602	.0766272	9.62	0.000	.5870829	.8878375

58 . mi estimate: regress dv_behavior c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_scale i.
> n##c.polengage_scale c.income##c.polengage_scale i.sample_num

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 970
Average RVI = 0.0100
Largest FMI = 0.0267
Complete DF = 951
DF adjustment: **Small sample** — DF: min = 899.21
avg = 935.95
max = 948.39
Model F test: **Equal FMI** — F(18, 948.7) = 4.11
Within VCE type: **OLS** — Prob > F = 0.0000

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	.0666241	.1464614	0.45	0.649	-.2208018	.3540501
polengage_scale	.2590463	.1818714	1.42	0.155	-.0978742	.6159668
c.latentnsc2#c.polengage_scale	-.234829	.1881441	-1.25	0.212	-.6040558	.1343978
age	.0901255	.1246461	0.72	0.470	-.1544888	.3347399
polengage_scale	0	(omitted)				
c.age#c.polengage_scale	-.1725505	.1629775	-1.06	0.290	-.4923898	.1472888
1.male	-.0707424	.0531278	-1.33	0.183	-.1750042	.0335193
polengage_scale	0	(omitted)				
male#c.polengage_scale						
1	.0172544	.0704212	0.25	0.806	-.1209454	.1554543
race						
1	-.047128	.1004187	-0.47	0.639	-.2441984	.1499423
2	-.0873595	.1275969	-0.68	0.494	-.3377691	.16305
3	-.0865041	.1229713	-0.70	0.482	-.3278407	.1548324
polengage_scale	0	(omitted)				
race#c.polengage_scale						
1	-.0058429	.1421805	-0.04	0.967	-.2848715	.2731856
2	.1136217	.1851794	0.61	0.540	-.2497945	.4770379
3	.0823609	.1745379	0.47	0.637	-.2601799	.4249018

education	.018429	.1131457	0.16	0.871	-.2036183	.2404763
0.polengage_scale	0	(omitted)				
c.education#c.polengage_scale	.0545586	.1532464	0.36	0.722	-.2461875	.3553048
income	.1224582	.0905643	1.35	0.177	-.0552838	.3002002
0.polengage_scale	0	(omitted)				
c.income#c.polengage_scale	-.0455863	.1195726	-0.38	0.703	-.2802562	.1890836
5.sample_num	-.0241642	.0163947	-1.47	0.141	-.0563385	.0080102
_cons	.7190468	.1297902	5.54	0.000	.464336	.9737575

```
59 .
60 .
   end of do-file

61 .
```

```
25 .
26 .
   end of do-file

27 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000006.tmp"

28 .
29 .
30 . ***** SM5 - Binning Estimator *****
31 .
32 . * Concern, Auth
33 . sum polengage_scale if dv_concern != . & auth_scale != . & polengage_scale < 0.33, det

                        polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%           0              0
 5%           0              0
10%           0              0       Obs                1,171
25%    .1269841              0       Sum of wgt.        1,171

50%          .2                      Mean            .174116
                       Largest       Std. dev.       .0947446
75%    .2430556       .3253968
90%    .2797619       .3253968       Variance        .0089765
95%    .3047619       .3253968       Skewness       -.7135387
99%    .3214286       .3263889       Kurtosis        2.488452

34 . sum polengage_scale if dv_concern != . & auth_scale != . & polengage_scale >= 0.33 & polengage_scale < 0.67, det

                        polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%    .3373016       .3333333
 5%    .3819444       .3333333
10%          .4       .3333333       Obs                3,933
25%    .4444444       .3333333       Sum of wgt.        3,933

50%    .5555556                      Mean            .5279618
                       Largest       Std. dev.       .0925026
75%          .6       .6666667
90%    .6388889       .6666667       Variance        .0085567
95%    .6626984       .6666667       Skewness       -.3345421
99%    .6666667       .6666667       Kurtosis        1.775385

35 . sum polengage_scale if dv_concern != . & auth_scale != . & polengage_scale >= 0.67, det // Bin Medians: 0.20, 0.56, 0.

                        polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%    .6904762       .6736111
 5%    .7321429       .6736111
10%    .7708333       .6736111       Obs               14,730
25%    .8333333       .6736111       Sum of wgt.       14,730

50%    .9285714                      Mean            .9047518
                       Largest       Std. dev.       .0944557
75%           1              1
90%           1              1       Variance        .0089219
95%           1              1       Skewness       -.5863582
99%           1              1       Kurtosis         2.13574
```

```
36 . mi estimate: regress dv_concern c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scale if
```

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 321
Average RVI = 0.0114
Largest FMI = 0.0985
Complete DF = 308
DF adjustment: **Small sample** — DF: min = 249.25
avg = 298.22
max = 305.64
Model F test: **Equal FMI** — F(12, 305.9) = 2.94
Within VCE type: **OLS** — Prob > F = 0.0007

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	.1545719	.0662175	2.33	0.020	.024272	.2848719
age	.0628502	.1082648	0.58	0.562	-.15019	.2758905
1.male	-.0267234	.0469989	-0.57	0.570	-.1192057	.065759
race						
1	-.0897568	.0666771	-1.35	0.179	-.2209625	.0414489
2	.0737711	.0764929	0.96	0.336	-.0767513	.2242935
3	.0172287	.0883662	0.19	0.846	-.1566625	.1911198
education	-.1019473	.0788424	-1.29	0.197	-.2570978	.0532031
income	-.1206192	.0953267	-1.27	0.207	-.3083678	.0671294
sample_num						
2	-.1830342	.0908525	-2.01	0.045	-.3618327	-.0042357
5	-.1373518	.1109929	-1.24	0.217	-.3557623	.0810586
6	-.1448987	.0664046	-2.18	0.030	-.2755896	-.0142077
polengage_scale	.1898569	.2133692	0.89	0.374	-.230006	.6097198
_cons	.6322168	.1043426	6.06	0.000	.4268814	.8375521

```
37 . mi estimate: regress dv_concern c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scale if
> e_scale < 0.67
```

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 1,358
Average RVI = 0.0051
Largest FMI = 0.0528
Complete DF = 1345
DF adjustment: **Small sample** — DF: min = 1,110.02
avg = 1,315.92
max = 1,342.68
Model F test: **Equal FMI** — F(12, 1342.8) = 8.39
Within VCE type: **OLS** — Prob > F = 0.0000

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.0780508	.0292231	-2.67	0.008	-.1353787	-.0207228
age	.1346551	.0430923	3.12	0.002	.0501196	.2191906
1.male	-.0479411	.0183518	-2.61	0.009	-.0839425	-.0119397
race						
1	-.1165058	.0371033	-3.14	0.002	-.1892927	-.043719
2	-.0019167	.0429825	-0.04	0.964	-.0862371	.0824037
3	-.0432205	.0460304	-0.94	0.348	-.13352	.0470791
education	-.0113253	.0341807	-0.33	0.740	-.0783796	.055729
income	-.0625503	.0408317	-1.53	0.126	-.1426664	.0175658
sample_num						

2	-.1489462	.0335281	-4.44	0.000	-.214721	-.0831714
5	-.1528585	.0337693	-4.53	0.000	-.2191051	-.086612
6	-.0362816	.0244122	-1.49	0.137	-.0841731	.0116098
polengage_scale	-.0540809	.0888236	-0.61	0.543	-.2283291	.1201672
_cons	.7865029	.0718117	10.95	0.000	.6456269	.9273788

```
38 . mi estimate: regress dv_concern c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scale if
```

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 3,805
Average RVI = 0.0124
Largest FMI = 0.1244
Complete DF = 3792
DF adjustment: Small sample — DF: min = 1,070.78
avg = 3,501.81
max = 3,787.30
Model F test: Equal FMI — F(12, 3780.9) = 31.74
Within VCE type: OLS — Prob > F = 0.0000

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.2174389	.0162343	-13.39	0.000	-.2492678	-.18561
age	-.0513212	.0286021	-1.79	0.073	-.1073981	.0047557
1.male	-.0597837	.0104753	-5.71	0.000	-.0803215	-.0392458
race						
1	-.0304102	.0225535	-1.35	0.178	-.0746286	.0138081
2	.0994962	.0291397	3.41	0.001	.0423649	.1566275
3	.015921	.0302606	0.53	0.599	-.0434079	.0752498
education	.0622771	.0200206	3.11	0.002	.0230242	.1015299
income	-.0626077	.0250008	-2.50	0.012	-.1116638	-.0135516
sample_num						
2	-.1526635	.0283086	-5.39	0.000	-.2081664	-.0971606
5	-.1947312	.0241031	-8.08	0.000	-.2419876	-.1474748
6	-.0817833	.0193946	-4.22	0.000	-.1198092	-.0437575
polengage_scale	.0325166	.0543424	0.60	0.550	-.0740266	.1390598
_cons	.8061459	.0566031	14.24	0.000	.69517	.9171218

```
39 .
40 . * Concern, NSC
41 . sum polengage_scale if dv_concern != . & latentnsc2 != . & polengage_scale < 0.50, det
```

polengage_scale

Percentiles		Smallest		
1%	.125	.0625		
5%	.125	.0625		
10%	.1875	.1180556	Obs	531
25%	.25	.125	Sum of wgt.	531
50%	.4097222		Mean	.3619612
	Largest		Std. dev.	.1102178
75%	.4513889	.4930556		
90%	.4652778	.4930556	Variance	.012148
95%	.4930556	.4930556	Skewness	-.8069601
99%	.4930556	.4930556	Kurtosis	2.349398

```
42 . sum polengage_scale if dv_concern != . & latentnsc2 != . & polengage_scale >= 0.50, det // Bin Medians: 0.41, 0.81

                      polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%           .5             .5
 5%     .5347222             .5
10%     .5763889             .5       Obs               1,384
25%     .7013889             .5       Sum of wgt.       1,384

50%        .8125                      Mean           .7983674
                       Largest       Std. dev.      .1422463
75%     .9166667              1
90%            1              1       Variance        .020234
95%            1              1       Skewness       -.4013471
99%            1              1       Kurtosis       2.204022

43 . mi estimate: regress dv_concern c.latentnsc2 c.age i.male i.race c.education c.income i.sample_num polengage_scale if

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        156
                                                Average RVI       =     0.0154
                                                Largest FMI       =     0.0352
                                                Complete DF       =        145
DF adjustment:   Small sample                   DF:     min       =     137.18
                                                        avg       =     140.98
                                                        max       =     142.50
Model F test:       Equal FMI                   F(  10,  142.9)   =       1.79
Within VCE type:          OLS                   Prob > F          =     0.0683

------------------------------------------------------------------------------
     dv_concern | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
----------------+-------------------------------------------------------------
     latentnsc2 |   .1424236   .1781449     0.80   0.425    -.2097247    .4945719
            age |   .3580354   .1480566     2.42   0.017     .0653584    .6507125
         1.male |   -.026487   .0574717    -0.46   0.646    -.1400947    .0871208
                |
           race |
             1  |  -.0913219   .1002331    -0.91   0.364    -.2894705    .1068267
             2  |     .21152   .1244222     1.70   0.091    -.0345134    .4575535
             3  |   .0042608   .1181932     0.04   0.971    -.2293983    .2379199
                |
      education |   .0457447     .13126     0.35   0.728    -.2137617    .3052511
         income |  -.0926013    .116744    -0.79   0.429     -.323445    .1382424
   5.sample_num |   .0428854   .0699749     0.61   0.541    -.0954459    .1812167
polengage_scale |   .1976746   .2716326     0.73   0.468    -.3392763    .7346255
          _cons |   .2818559   .1662738     1.70   0.092    -.0468312    .6105429
------------------------------------------------------------------------------

44 . quietly mimrgns, at(latentnsc2=( 0.055  0.782)) coeflegend post

45 . lincom _b[2._at]-_b[1._at]

 ( 1)  - 1bn._at + 2._at = 0

------------------------------------------------------------------------------
             | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
         (1) |   .1035419   .1295114     0.80   0.425    -.1524321    .3595159
------------------------------------------------------------------------------
```

```
46 . mi estimate: regress dv_concern c.latentnsc2 c.age i.male i.race c.education c.income i.sample_num polengage_scale if

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        809
                                                Average RVI       =     0.0053
                                                Largest FMI       =     0.0211
                                                Complete DF       =        798
DF adjustment:   Small sample                   DF:     min       =     768.30
                                                        avg       =     788.96
                                                        max       =     795.81
Model F test:       Equal FMI                   F(  10,  795.9)   =       3.55
Within VCE type:          OLS                   Prob > F          =     0.0001
```

dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.1307543	.060883	-2.15	0.032	-.2502645	-.0112441
age	-.0402295	.0593406	-0.68	0.498	-.1567122	.0762532
1.male	-.0574025	.0249225	-2.30	0.022	-.1063242	-.0084809
race						
1	-.0413059	.0543796	-0.76	0.448	-.1480528	.065441
2	.0686857	.0677075	1.01	0.311	-.064228	.2015994
3	.0367298	.0662559	0.55	0.579	-.0933323	.1667919
education	.1245666	.0553289	2.25	0.025	.0159582	.233175
income	-.092968	.0462358	-2.01	0.045	-.1837275	-.0022085
5.sample_num	-.0333388	.0284822	-1.17	0.242	-.089248	.0225704
polengage_scale	.2255368	.0908466	2.48	0.013	.0472093	.4038643
_cons	.4244843	.0930554	4.56	0.000	.2418196	.607149

```
47 . quietly mimrgns, at(latentnsc2=( 0.055  0.782)) coeflegend post

48 . lincom _b[2._at]-_b[1._at]

 ( 1)  - 1bn._at + 2._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0950584	.0442619	-2.15	0.032	-.1819419	-.0081748

```
49 .
50 . * Restrictions, Auth Medians: 0.20, 0.52, 0.93
51 . sum polengage_scale if dv_restrictions != . & auth_scale != . & polengage_scale < 0.33, det

                      polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%            0              0
 5%            0              0
10%            0              0       Obs                2,394
25%           .1              0       Sum of wgt.        2,394

50%     .1982143                      Mean           .1820067
                        Largest       Std. dev.      .1058327
75%     .2678571        .327381
90%     .3005952        .327381       Variance       .0112006
95%     .3214286        .327381       Skewness       -.526842
99%     .3214286        .327381       Kurtosis       2.027336
```

```
52 . sum polengage_scale if dv_restrictions != . & auth_scale != . & polengage_scale >= 0.33 & polengage_scale < 0.67, det

                      polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%     .3321429       .3303571
 5%     .3607143       .3303571
10%     .3839286       .3321429       Obs               4,397
25%     .4410714       .3321429       Sum of wgt.       4,397

50%     .5214286                      Mean           .5181626
                       Largest       Std. dev.      .0963672
75%           .6       .6696429
90%     .6517857       .6696429       Variance       .0092866
95%     .6626984       .6696429       Skewness      -.1850394
99%     .6678571       .6696429       Kurtosis       1.942981

53 . sum polengage_scale if dv_restrictions != . & auth_scale != . & polengage_scale >= 0.67, det

                      polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%     .6785714        .672619
 5%     .7142857        .672619
10%          .75        .672619       Obs              15,029
25%     .8333333        .672619       Sum of wgt.      15,029

50%     .9285714                      Mean           .9022052
                       Largest       Std. dev.      .0978961
75%            1              1
90%            1              1       Variance       .0095836
95%            1              1       Skewness      -.6422863
99%            1              1       Kurtosis         2.2129

54 .
55 . mi estimate: regress dv_restrictions c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scal

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        694
                                                Average RVI       =     0.0249
                                                Largest FMI       =     0.1188
                                                Complete DF       =        682
DF adjustment:   Small sample                   DF:     min       =     446.31
                                                        avg       =     634.68
                                                        max       =     679.48
Model F test:       Equal FMI                   F(  11,  678.6)   =       8.04
Within VCE type:          OLS                   Prob > F          =     0.0000

------------------------------------------------------------------------------
dv_restrictions | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
----------------+-------------------------------------------------------------
     auth_scale |   .0185228   .0336297     0.55   0.582     -.047508    .0845536
            age |  -.0283756   .0550231    -0.52   0.606    -.1365121    .0797609
         1.male |    .021137   .0224702     0.94   0.347    -.0229933    .0652673
                |
           race |
             1  |  -.0401345    .034501    -1.16   0.245    -.1078761    .0276071
             2  |   .0295374   .0407933     0.72   0.469    -.0505587    .1096336
             3  |   .0410745   .0409168     1.00   0.316    -.0392641    .1214131
                |
      education |  -.0030004   .0422118    -0.07   0.943    -.0858863    .0798855
         income |   .0291458   .0450333     0.65   0.518    -.0593272    .1176188
                |
     sample_num |
             4  |  -.0912481   .0289393    -3.15   0.002    -.1480694   -.0344269
             6  |   .1086809   .0353629     3.07   0.002     .0392424    .1781194
------------------------------------------------------------------------------
```

polengage_scale	-.0129033	.1043094	-0.12	0.902	-.2177112	.1919046
_cons	.5718867	.0535737	10.67	0.000	.4666958	.6770776

56 . mi estimate: regress dv_restrictions c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scal
> engage_scale < 0.67

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 1,947
Average RVI = 0.0052
Largest FMI = 0.0313
Complete DF = 1935
DF adjustment: Small sample — DF: min = 1,740.82
avg = 1,906.38
max = 1,932.47
Model F test: Equal FMI — F(11, 1932.5) = 15.39
Within VCE type: OLS — Prob > F = 0.0000

dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.0292868	.0200866	-1.46	0.145	-.0686806	.010107
age	-.1167161	.0307625	-3.79	0.000	-.1770484	-.0563837
1.male	.0002607	.0129364	0.02	0.984	-.0251102	.0256317
race						
1	-.0356347	.0227987	-1.56	0.118	-.0803474	.0090781
2	.0560834	.0273859	2.05	0.041	.0023743	.1097924
3	.0362538	.0272093	1.33	0.183	-.0171088	.0896165
education	.0734303	.0249679	2.94	0.003	.0244631	.1223974
income	.068159	.0252045	2.70	0.007	.0187247	.1175932
sample_num						
4	-.0501596	.0162461	-3.09	0.002	-.0820214	-.0182979
6	.1022314	.017694	5.78	0.000	.0675301	.1369328
polengage_scale	.0857398	.0616924	1.39	0.165	-.0352509	.2067305
_cons	.5747358	.0438142	13.12	0.000	.4888076	.6606639

57 . mi estimate: regress dv_restrictions c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scal

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 4,229
Average RVI = 0.0107
Largest FMI = 0.0944
Complete DF = 4217
DF adjustment: Small sample — DF: min = 1,592.85
avg = 3,943.33
max = 4,214.36
Model F test: Equal FMI — F(11, 4205.8) = 55.13
Within VCE type: OLS — Prob > F = 0.0000

dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.2437809	.0144268	-16.90	0.000	-.2720651	-.2154967
age	-.1633948	.0260886	-6.26	0.000	-.2145429	-.1122467
1.male	-.0589184	.0098429	-5.99	0.000	-.0782158	-.039621
race						
1	-.0193967	.0195216	-0.99	0.320	-.0576695	.018876
2	.1769011	.0255991	6.91	0.000	.1267135	.2270888
3	.0319361	.0260811	1.22	0.221	-.0191965	.0830688
education	.0938421	.0185099	5.07	0.000	.0575523	.130132

income	.0230064	.0208989	1.10	0.271	-.0179859	.0639986
sample_num						
4	-.0923043	.0169428	-5.45	0.000	-.1255212	-.0590875
6	-.0286525	.0149307	-1.92	0.055	-.0579245	.0006195
polengage_scale	-.1582849	.0496012	-3.19	0.001	-.2555293	-.0610405
_cons	.9485401	.0489874	19.36	0.000	.8524989	1.044581

```
58 .
59 . * Behavior, Auth Medians: 0.22, 0.50, 0.84
60 . sum polengage_scale if dv_behavior != . & auth_scale != . & polengage_scale < 0.33, det
```

polengage_scale

Percentiles		Smallest		
1%	0	0		
5%	0	0		
10%	0	0	Obs	1,200
25%	.1	0	Sum of wgt.	1,200
50%	.2222222		Mean	.1910513
		Largest	Std. dev.	.1018395
75%	.2678571	.327381		
90%	.3	.327381	Variance	.0103713
95%	.3214286	.327381	Skewness	-.6396216
99%	.3214286	.327381	Kurtosis	2.218641

```
61 . sum polengage_scale if dv_behavior != . & auth_scale != . & polengage_scale >= 0.33 & polengage_scale < 0.67, det
```

polengage_scale

Percentiles		Smallest		
1%	.3321429	.3303571		
5%	.3410714	.3303571		
10%	.3660714	.3321429	Obs	2,392
25%	.4166667	.3321429	Sum of wgt.	2,392
50%	.4964286		Mean	.4964411
		Largest	Std. dev.	.0976936
75%	.5833333	.6696429		
90%	.6321429	.6696429	Variance	.009544
95%	.6607143	.6696429	Skewness	.101454
99%	.6678571	.6696429	Kurtosis	1.864544

```
62 . sum polengage_scale if dv_behavior != . & auth_scale != . & polengage_scale >= 0.67, det
```

polengage_scale

Percentiles		Smallest		
1%	.6736111	.672619		
5%	.6845238	.672619		
10%	.7017857	.672619	Obs	2,962
25%	.7660714	.672619	Sum of wgt.	2,962
50%	.8375		Mean	.8446495
		Largest	Std. dev.	.1018105
75%	.9375	1		
90%	1	1	Variance	.0103654
95%	1	1	Skewness	.0749979
99%	1	1	Kurtosis	1.822397

```
63 .
64 . mi estimate: regress dv_behavior c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scale it

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        500
                                                Average RVI       =     0.0152
                                                Largest FMI       =     0.0686
                                                Complete DF       =        487
DF adjustment:   Small sample                   DF:     min       =     415.93
                                                        avg       =     471.85
                                                        max       =     484.61
Model F test:       Equal FMI                   F(  12,  484.7)   =      11.48
Within VCE type:          OLS                   Prob > F          =     0.0000

------------------------------------------------------------------------------
 dv_behavior | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
  auth_scale |  -.0002055     .04156    -0.00   0.996    -.0818664    .0814553
         age |   .2391868     .05834     4.10   0.000     .1245305     .353843
      1.male |  -.0204459   .0249703    -0.82   0.413    -.0695296    .0286378
             |
        race |
          1  |  -.0304337    .042377    -0.72   0.473    -.1136994    .0528321
          2  |   .0942259   .0495318     1.90   0.058    -.0030979    .1915498
          3  |   .0921116   .0472905     1.95   0.052    -.0008087    .1850319
             |
   education |  -.0014169    .049723    -0.03   0.977    -.0991187     .096285
      income |    .085696   .0481742     1.78   0.076    -.0089843    .1803763
             |
  sample_num |
          3  |   .0911261    .058648     1.55   0.121    -.0241098    .2063619
          4  |  -.1623136   .0540713    -3.00   0.003    -.2685567   -.0560704
          5  |   .1087021   .0910861     1.19   0.233    -.0702703    .2876745
             |
polengage_scale |   .1297274   .1235645     1.05   0.294    -.1130628    .3725176
       _cons |   .5979349   .0748878     7.98   0.000     .4507891    .7450806
------------------------------------------------------------------------------

65 . mi estimate: regress dv_behavior c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scale it
   > ge_scale < 0.67

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      1,567
                                                Average RVI       =     0.0044
                                                Largest FMI       =     0.0253
                                                Complete DF       =       1554
DF adjustment:   Small sample                   DF:     min       =   1,454.36
                                                        avg       =   1,536.36
                                                        max       =   1,550.84
Model F test:       Equal FMI                   F(  12, 1551.8)   =      19.96
Within VCE type:          OLS                   Prob > F          =     0.0000

------------------------------------------------------------------------------
 dv_behavior | Coefficient  Std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
  auth_scale |  -.0183014   .0229363    -0.80   0.425    -.0632909    .0266882
         age |   .0874976   .0306222     2.86   0.004      .027431    .1475642
      1.male |  -.0180481   .0128153    -1.41   0.159    -.0431858    .0070895
             |
        race |
          1  |   -.063352    .022632    -2.80   0.005    -.1077446   -.0189594
          2  |   .0024909   .0273499     0.09   0.927    -.0511559    .0561377
          3  |   .0044744   .0263233     0.17   0.865    -.0471588    .0561075
             |
   education |   .0929156   .0258769     3.59   0.000     .0421579    .1436733
      income |   .0791191   .0238568     3.32   0.001     .0323217    .1259165
```

sample_num						
3	.0039011	.0219755	0.18	0.859	-.0392039	.047006
4	-.1468055	.0197889	-7.42	0.000	-.1856213	-.1079897
5	-.0175676	.0273778	-0.64	0.521	-.071269	.0361337
polengage_scale	.1624891	.0636147	2.55	0.011	.0377091	.287269
_cons	.6809246	.0442363	15.39	0.000	.5941554	.7676938

66 . mi estimate: regress dv_behavior c.auth_scale c.age i.male i.race c.education c.income i.sample_num polengage_scale it

Multiple-imputation estimates		Imputations	= 25
Linear regression		Number of obs	= 1,962
		Average RVI	= 0.0055
		Largest FMI	= 0.0219
		Complete DF	= 1949
DF adjustment: Small sample		DF: min	= 1,835.21
		avg	= 1,926.31
		max	= 1,945.76
Model F test: Equal FMI		F(12, 1946.5)	= 27.31
Within VCE type: OLS		Prob > F	= 0.0000

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
auth_scale	-.0639393	.0173106	-3.69	0.000	-.0978886	-.02999
age	-.0353325	.0247792	-1.43	0.154	-.0839308	.0132658
1.male	-.0260008	.0107648	-2.42	0.016	-.0471128	-.0048889
race						
1	-.0188947	.0209356	-0.90	0.367	-.0599535	.0221641
2	.0743745	.0273491	2.72	0.007	.0207376	.1280113
3	.0129992	.02607	0.50	0.618	-.0381293	.0641276
education	.0533869	.0215892	2.47	0.013	.0110463	.0957276
income	.0729594	.0190641	3.83	0.000	.0355706	.1103482
sample_num						
3	.0050978	.0180688	0.28	0.778	-.0303384	.0405341
4	-.155424	.016653	-9.33	0.000	-.1880836	-.1227643
5	-.0099074	.0195312	-0.51	0.612	-.0482117	.028397
polengage_scale	-.0325518	.0539886	-0.60	0.547	-.1384335	.07333
_cons	.8860703	.0506215	17.50	0.000	.7867921	.9853485

67 .
68 . * Behavior, NSC Medians: 0.41, 0.81
69 . sum polengage_scale if dv_concern != . & latentnsc2 != . & polengage_scale < 0.50, det

polengage_scale

Percentiles	Smallest		
1% .125	.0625		
5% .125	.0625		
10% .1875	.1180556	Obs	531
25% .25	.125	Sum of wgt.	531
50% .4097222		Mean	.3619612
	Largest	Std. dev.	.1102178
75% .4513889	.4930556		
90% .4652778	.4930556	Variance	.012148
95% .4930556	.4930556	Skewness	-.8069601
99% .4930556	.4930556	Kurtosis	2.349398

```
70 . sum polengage_scale if dv_concern != . & latentnsc2 != . & polengage_scale >= 0.50, det

                      polengage_scale
-------------------------------------------------------------
      Percentiles      Smallest
 1%           .5             .5
 5%     .5347222             .5
10%     .5763889             .5       Obs               1,384
25%     .7013889             .5       Sum of wgt.       1,384

50%        .8125                     Mean           .7983674
                       Largest       Std. dev.      .1422463
75%     .9166667             1
90%            1             1       Variance        .020234
95%            1             1       Skewness       -.4013471
99%            1             1       Kurtosis        2.204022

71 .
72 . mi estimate: regress dv_behavior c.latentnsc2 c.age i.male i.race c.education c.income i.sample_num polengage_scale if
```

```
Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        159
                                                Average RVI       =     0.0092
                                                Largest FMI       =     0.0381
                                                Complete DF       =        148
DF adjustment:   Small sample                   DF:     min       =     139.46
                                                        avg       =     144.53
                                                        max       =     146.00
Model F test:       Equal FMI                   F(  10,  146.0)   =       1.02
Within VCE type:          OLS                   Prob > F          =     0.4317
```

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.0462319	.1248324	-0.37	0.712	-.2929457	.2004819
age	.14902	.1052718	1.42	0.159	-.0590382	.3570781
1.male	-.049027	.0417698	-1.17	0.242	-.1315796	.0335256
race						
1	-.1139767	.0714286	-1.60	0.113	-.255151	.0271976
2	-.0695361	.0874767	-0.79	0.428	-.242431	.1033587
3	-.0901202	.088984	-1.01	0.313	-.2660204	.08578
education	.0317436	.0973452	0.33	0.745	-.1606619	.2241491
income	.0852873	.0825596	1.03	0.303	-.0779429	.2485175
5.sample_num	-.0126039	.0509734	-0.25	0.805	-.1133538	.0881461
polengage_scale	.2957082	.2028774	1.46	0.147	-.1052478	.6966641
_cons	.7286518	.1242148	5.87	0.000	.4831563	.9741472

```
73 . quietly mimrgns, at(latentnsc2=( 0.055  0.782)) coeflegend post

74 . lincom _b[2._at]-_b[1._at]

 ( 1)  - 1bn._at + 2._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0336106	.0907531	-0.37	0.712	-.2129499	.1457287

```
75 . mi estimate: regress dv_behavior c.latentnsc2 c.age i.male i.race c.education c.income i.sample_num polengage_scale it

   Multiple-imputation estimates                   Imputations       =         25
   Linear regression                               Number of obs     =        811
                                                   Average RVI       =     0.0109
                                                   Largest FMI       =     0.0239
                                                   Complete DF       =        800
   DF adjustment:   Small sample                   DF:     min       =     764.95
                                                           avg       =     785.70
                                                           max       =     796.96
   Model F test:       Equal FMI                   F(  10,  797.6)   =       5.80
   Within VCE type:          OLS                   Prob > F          =     0.0000
```

dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
latentnsc2	-.1156731	.0363723	-3.18	0.002	-.18707	-.0442762
age	-.0586833	.0352339	-1.67	0.096	-.1278467	.0104802
1.male	-.0597872	.0150016	-3.99	0.000	-.0892349	-.0303394
race						
1	-.034304	.0324264	-1.06	0.290	-.097958	.0293499
2	.0036143	.0401592	0.09	0.928	-.0752182	.0824468
3	-.0130484	.0397123	-0.33	0.743	-.0910065	.0649096
education	.0571781	.0339168	1.69	0.092	-.0094007	.1237569
income	.0885325	.0283782	3.12	0.002	.0328263	.1442387
5.sample_num	-.0251044	.0171375	-1.46	0.143	-.0587448	.0085359
polengage_scale	.0998823	.0545009	1.83	0.067	-.0071	.2068646
_cons	.8367152	.0561109	14.91	0.000	.7265715	.9468588

```
76 . quietly mimrgns, at(latentnsc2=( 0.055  0.782)) coeflegend post

77 . lincom _b[2._at]-_b[1._at]

    ( 1)  - 1bn._at + 2._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0840943	.0264427	-3.18	0.002	-.1359996	-.0321891

```
78 .
79 .
80 .
   end of do-file

81 .
```

```
20 . ***** SM7 - Political Engagement vs. Interest vs. Knowledge *****
21 .
```

```
22 . *Concern, Authoritarianism
23 . mi estimate: regress z_dv_concern c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
   >  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =      5,484
                                              Average RVI       =     0.0037
                                              Largest FMI       =     0.1110
                                              Complete DF       =       5412
DF adjustment:   Small sample                 DF:     min       =   1,402.93
                                                      avg       =   5,221.73
                                                      max       =   5,406.85
Model F test:       Equal FMI                 F(  71, 5409.7)   =       8.30
Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1288974	.0144133	-8.94	0.000	-.1571533	-.1006415
z_polengage_scale	.0529836	.0706384	0.75	0.453	-.0854963	.1914635
c.z_auth_scale#c.z_polengage_scale	-.1059967	.0147232	-7.20	0.000	-.13486	-.0771334
age	.0422055	.0707845	0.60	0.551	-.0965606	.1809715
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.2195633	.0737417	-2.98	0.003	-.3641267	-.0749999
1.male	-.1590707	.0289746	-5.49	0.000	-.2158726	-.1022687
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.0618112	.0303392	-2.04	0.042	-.1212883	-.0023341
race						
1	-.1881466	.05789	-3.25	0.001	-.3016345	-.0746587
2	.1987479	.0711719	2.79	0.005	.0592219	.3382738
3	.0051046	.0744291	0.07	0.945	-.1408068	.1510159
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0923227	.0577384	1.60	0.110	-.020868	.2055135
2	.0701478	.0686927	1.02	0.307	-.0645178	.2048134
3	.0264831	.0736845	0.36	0.719	-.1179684	.1709347
education	.0688295	.0539728	1.28	0.202	-.0369806	.1746396
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0996942	.05437	1.83	0.067	-.0068935	.2062818
income	-.2035268	.0659163	-3.09	0.002	-.3328319	-.0742217
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	.0015558	.0583848	0.03	0.979	-.1129251	.1160367
sample_num						
2	-.4664103	.0639281	-7.30	0.000	-.5917389	-.3410816
5	-.552779	.0576996	-9.58	0.000	-.665894	-.4396641
6	-.2286309	.0447661	-5.11	0.000	-.3163937	-.140868
state_1	.016512	.4878308	0.03	0.973	-.9398356	.9728595
state_2	-.1029886	.4455278	-0.23	0.817	-.976406	.7704287
state_3	.2232482	.4474741	0.50	0.618	-.6539844	1.100481
state_4	-.2547854	.4339538	-0.59	0.557	-1.105513	.5959423
state_5	.0548061	.4282113	0.13	0.898	-.784664	.8942761
state_6	.025127	.4382939	0.06	0.954	-.8341085	.8843626

state_7	.0851541	.4517825	0.19	0.851	-.8005247	.970833
state_8	.5200265	.5202917	1.00	0.318	-.4999571	1.54001
state_9	.2534801	.4729826	0.54	0.592	-.6737591	1.180719
state_10	.022424	.4285428	0.05	0.958	-.817696	.8625439
state_11	-.0012524	.4317666	-0.00	0.998	-.8476924	.8451876
state_12	.2235539	.4816003	0.46	0.643	-.7205794	1.167687
state_13	.1532342	.4399469	0.35	0.728	-.7092423	1.015711
state_14	-.0338892	.4507765	-0.08	0.940	-.9175957	.8498173
state_15	.1048642	.4311142	0.24	0.808	-.7402967	.950025
state_16	-.0245844	.4363272	-0.06	0.955	-.8799647	.830796
state_17	.0903309	.4443526	0.20	0.839	-.7807823	.9614441
state_18	.1894995	.4431431	0.43	0.669	-.6792428	1.058242
state_19	-.1140272	.4508259	-0.25	0.800	-.9978309	.7697766
state_20	.0486416	.4368377	0.11	0.911	-.8077394	.9050226
state_21	.2922445	.4362157	0.67	0.503	-.5629172	1.147406
state_22	.2458594	.4606557	0.53	0.594	-.6572142	1.148933
state_23	-.0973335	.4316267	-0.23	0.822	-.9434991	.7488322
state_24	-.0732644	.4396339	-0.17	0.868	-.9351274	.7885985
state_25	-.0752477	.4360905	-0.17	0.863	-.9301642	.7796688
state_26	.1642978	.4515443	0.36	0.716	-.7209138	1.049509
state_27	-.1484153	.477867	-0.31	0.756	-1.08523	.7883995
state_28	-.0036848	.4329666	-0.01	0.993	-.8524772	.8451077
state_29	-.5448718	.5991074	-0.91	0.363	-1.719365	.6296218
state_30	-.0048529	.4643601	-0.01	0.992	-.9151886	.9054827
state_31	.0364696	.4644939	0.08	0.937	-.8741285	.9470677
state_32	.0798894	.4334373	0.18	0.854	-.7698255	.9296044
state_33	-.172503	.4572193	-0.38	0.706	-1.06884	.7238338
state_34	.0014553	.4467537	0.00	0.997	-.8743651	.8772756
state_35	.1740861	.4295023	0.41	0.685	-.6679148	1.016087
state_36	.136276	.4316399	0.32	0.752	-.7099154	.9824674
state_37	.2964637	.4528901	0.65	0.513	-.5913867	1.184314
state_38	-.1519864	.4412211	-0.34	0.731	-1.016961	.7129881
state_39	.0832983	.4299506	0.19	0.846	-.7595816	.9261781
state_40	.3123403	.4870922	0.64	0.521	-.6425594	1.26724
state_41	.0252984	.4419353	0.06	0.954	-.8410762	.891673
state_42	.1805261	.4735323	0.38	0.703	-.7477909	1.108843
state_43	.0047964	.4370359	0.01	0.991	-.8519733	.8615662
state_44	-.0299898	.4289037	-0.07	0.944	-.8708172	.8108377
state_45	-.0811312	.4471199	-0.18	0.856	-.9576697	.7954073
state_46	.0196604	.434427	0.05	0.964	-.8319947	.8713156
state_47	.3265047	.5002002	0.65	0.514	-.6540913	1.307101
state_48	-.1488706	.4336835	-0.34	0.731	-.9990682	.701327
state_49	.0336234	.4350129	0.08	0.938	-.8191805	.8864274
state_50	.1696	.4520241	0.38	0.708	-.7165526	1.055753
state_51	-.0888351	.5202316	-0.17	0.864	-1.108701	.9310309
_cons	.4285897	.4347809	0.99	0.324	-.4237609	1.28094

```
24 . mi estimate: regress z_dv_concern c.z_auth_scale##c.z_polint_scale c.age##c.z_polint_scale i.male##c.z_polint_scale i.
> ##c.z_polint_scale c.income##c.z_polint_scale i.sample_num state_1-state_51

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      4,639
                                                Average RVI       =     0.0042
                                                Largest FMI       =     0.0788
                                                Complete DF       =       4568
DF adjustment:   Small sample                   DF:     min       =   2,028.94
                                                        avg       =   4,302.66
                                                        max       =   4,560.95
Model F test:       Equal FMI                   F(  70, 4565.8)   =       7.30
Within VCE type:          OLS                   Prob > F          =     0.0000
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1215605	.0165803	-7.33	0.000	-.1540659	-.0890552
z_polint_scale	.1347436	.0849468	1.59	0.113	-.0317937	.3012808
c.z_auth_scale#c.z_polint_scale	-.1193149	.0178949	-6.67	0.000	-.1543975	-.0842322
age	-.1483899	.0862527	-1.72	0.085	-.3174869	.0207071
z_polint_scale	0	(omitted)				
c.age#c.z_polint_scale	-.2387683	.0955307	-2.50	0.012	-.4260547	-.0514819
1.male	-.1832254	.0327241	-5.60	0.000	-.2473805	-.1190702
z_polint_scale	0	(omitted)				
male#c.z_polint_scale						
1	-.0557688	.0362082	-1.54	0.124	-.1267544	.0152168
race						
1	-.1786006	.0635772	-2.81	0.005	-.3032427	-.0539585
2	.1649607	.0788967	2.09	0.037	.0102846	.3196368
3	.0065882	.0837074	0.08	0.937	-.157519	.1706955
z_polint_scale	0	(omitted)				
race#c.z_polint_scale						
1	.0535145	.0669773	0.80	0.424	-.0777936	.1848226
2	.1530787	.0832508	1.84	0.066	-.0101338	.3162913
3	.0552236	.0881873	0.63	0.531	-.1176668	.2281139
education	.0934006	.0633832	1.47	0.141	-.0308625	.2176637
z_polint_scale	0	(omitted)				
c.education#c.z_polint_scale	.0473766	.0690578	0.69	0.493	-.0880113	.1827646
income	-.2209395	.0720619	-3.07	0.002	-.3622625	-.0796164
z_polint_scale	0	(omitted)				
c.income#c.z_polint_scale	.0334215	.0765874	0.44	0.663	-.116769	.1836119
sample_num						
5	-.0916447	.0684128	-1.34	0.180	-.2257671	.0424776
6	.2101529	.0530266	3.96	0.000	.1061951	.3141107
state_1	-.1258353	.6055933	-0.21	0.835	-1.313107	1.061436
state_2	-.2135826	.5726295	-0.37	0.709	-1.336231	.9090659
state_3	.0908327	.5731357	0.16	0.874	-1.032808	1.214474
state_4	-.446504	.5625589	-0.79	0.427	-1.54941	.6564022
state_5	-.0902075	.5581455	-0.16	0.872	-1.184461	1.004046
state_6	-.1597229	.5660249	-0.28	0.778	-1.269423	.9499769
state_7	-.1063274	.5778352	-0.18	0.854	-1.239181	1.026527
state_8	.4211993	.632825	0.67	0.506	-.819457	1.661856
state_9	.1186114	.5937741	0.20	0.842	-1.045489	1.282712
state_10	-.1101629	.5582111	-0.20	0.844	-1.204545	.9842193
state_11	-.1289648	.5608283	-0.23	0.818	-1.228478	.9705486
state_12	.082914	.6029626	0.14	0.891	-1.0992	1.265027
state_13	.0302325	.5673082	0.05	0.958	-1.081984	1.142449
state_14	-.272534	.5762949	-0.47	0.636	-1.402368	.8572999
state_15	.0054576	.5603908	0.01	0.992	-1.093198	1.104113
state_16	-.1356904	.5646297	-0.24	0.810	-1.242656	.971275
state_17	.0057586	.5711971	0.01	0.992	-1.114082	1.125599
state_18	.0387858	.5700842	0.07	0.946	-1.078873	1.156445
state_19	-.2423933	.5764627	-0.42	0.674	-1.372557	.8877706
state_20	-.098915	.5651876	-0.18	0.861	-1.206974	1.009144
state_21	.1695884	.5644005	0.30	0.764	-.9369275	1.276104
state_22	.1029263	.5836925	0.18	0.860	-1.04141	1.247263

state_23	-.2126327	.5609734	-0.38	0.705	-1.31243	.8871649
state_24	-.1946282	.5673494	-0.34	0.732	-1.306926	.9176693
state_25	-.1994563	.5645264	-0.35	0.724	-1.30622	.9073078
state_26	-.0094992	.5777403	-0.02	0.987	-1.142166	1.123167
state_27	-.3636097	.5976505	-0.61	0.543	-1.53531	.8080909
state_28	-.1207224	.5618959	-0.21	0.830	-1.222329	.9808847
state_29	-.6681977	.7008462	-0.95	0.340	-2.042206	.705811
state_30	-.1245494	.5868018	-0.21	0.832	-1.274981	1.025882
state_31	-.1127149	.5868788	-0.19	0.848	-1.263297	1.037868
state_32	-.0750182	.5621557	-0.13	0.894	-1.177133	1.027097
state_33	-.4044758	.5813245	-0.70	0.487	-1.544169	.7352173
state_34	-.1723987	.5725977	-0.30	0.763	-1.294985	.9501875
state_35	.0134295	.5590786	0.02	0.981	-1.082654	1.109513
state_36	.0343304	.5607976	0.06	0.951	-1.065123	1.133784
state_37	.1629109	.5775435	0.28	0.778	-.969372	1.295194
state_38	-.3052322	.5681302	-0.54	0.591	-1.41906	.808596
state_39	-.0671068	.5593436	-0.12	0.905	-1.16371	1.029496
state_40	.1325502	.6051341	0.22	0.827	-1.053822	1.318922
state_41	-.1163152	.5690881	-0.20	0.838	-1.232022	.9993913
state_42	.044235	.5941784	0.07	0.941	-1.120659	1.209129
state_43	-.1281961	.5652484	-0.23	0.821	-1.236375	.9799827
state_44	-.1721077	.5586616	-0.31	0.758	-1.267373	.9231577
state_45	-.261699	.5733894	-0.46	0.648	-1.385838	.8624396
state_46	-.107416	.5631021	-0.19	0.849	-1.211387	.9965546
state_47	.1667004	.6160984	0.27	0.787	-1.041164	1.374565
state_48	-.2669555	.56223	-0.47	0.635	-1.369216	.8353051
state_49	-.0829835	.5634756	-0.15	0.883	-1.187687	1.02172
state_50	.0832664	.5772247	0.14	0.885	-1.048391	1.214924
state_51	-.2585879	.6329193	-0.41	0.683	-1.49943	.9822543
_cons	.2036026	.5628559	0.36	0.718	-.8998874	1.307093

25 . mi estimate: regress z_dv_concern c.z_auth_scale##c.z_polknow_scale c.age##c.z_polknow_scale i.male##c.z_polknow_scale
> tion##c.z_polknow_scale c.income##c.z_polknow_scale i.sample_num state_1-state_51

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 5,154
Average RVI = 0.0032
Largest FMI = 0.1166
Complete DF = 5082
DF adjustment: **Small sample** — DF: min = 1,283.97
avg = 4,928.58
max = 5,076.66
Model F test: **Equal FMI** — F(71, 5079.8) = 7.49
Within VCE type: **OLS** — Prob > F = 0.0000

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1489365	.0143727	-10.36	0.000	-.1771132	-.1207599
z_polknow_scale	-.0521381	.0729076	-0.72	0.475	-.1950685	.0907923
c.z_auth_scale#c.z_polknow_scale	-.0844824	.0148529	-5.69	0.000	-.1136006	-.0553642
age	.0374826	.0729149	0.51	0.607	-.1054621	.1804273
z_polknow_scale	0	(omitted)				
c.age#c.z_polknow_scale	-.1425565	.0760483	-1.87	0.061	-.2916439	.0065309
1.male	-.1397054	.02928	-4.77	0.000	-.1971069	-.0823039
z_polknow_scale	0	(omitted)				
male#c.z_polknow_scale						
1	-.0574482	.0310913	-1.85	0.065	-.1184005	.0035042
race						
1	-.1545334	.0590057	-2.62	0.009	-.2702102	-.0388567

2	.2180149	.0742131	2.94	0.003	.0725251	.3635046
3	.0303317	.0764085	0.40	0.691	-.119462	.1801254
z_polknow_scale	0	(omitted)				
race#c.z_polknow_scale						
1	.091734	.0582626	1.57	0.115	-.0224858	.2059539
2	.0439866	.0708947	0.62	0.535	-.0949976	.1829707
3	.0356424	.0764214	0.47	0.641	-.1141767	.1854616
education	.0967618	.0544031	1.78	0.075	-.0098937	.2034173
z_polknow_scale	0	(omitted)				
c.education#c.z_polknow_scale	.1128157	.0555069	2.03	0.042	.0039977	.2216337
income	-.1901721	.0678056	-2.80	0.005	-.3231941	-.0571501
z_polknow_scale	0	(omitted)				
c.income#c.z_polknow_scale	.0295453	.0603179	0.49	0.624	-.0887236	.1478141
sample_num						
2	-.4964071	.0657754	-7.55	0.000	-.6253601	-.3674542
5	-.5736107	.0593726	-9.66	0.000	-.6900072	-.4572142
6	-.1925098	.0445983	-4.32	0.000	-.2799456	-.1050741
state_1	.003902	.5027805	0.01	0.994	-.9817664	.9895705
state_2	-.1725939	.4483573	-0.38	0.700	-1.05157	.706382
state_3	.2110245	.4510411	0.47	0.640	-.6732127	1.095262
state_4	-.3424216	.4366805	-0.78	0.433	-1.198506	.5136628
state_5	-.014323	.4299669	-0.03	0.973	-.857246	.8285999
state_6	.0284579	.4405707	0.06	0.949	-.8352529	.8921686
state_7	.0441362	.4554675	0.10	0.923	-.8487786	.937051
state_8	.4492527	.5221335	0.86	0.390	-.5743556	1.472861
state_9	.1990738	.4747147	0.42	0.675	-.7315737	1.129721
state_10	-.0412729	.4302621	-0.10	0.924	-.8847746	.8022288
state_11	-.0113052	.4338751	-0.03	0.979	-.86189	.8392796
state_12	.2307277	.4866755	0.47	0.635	-.7233679	1.184823
state_13	.0837862	.441752	0.19	0.850	-.7822406	.9498129
state_14	-.0567699	.4543195	-0.12	0.901	-.947434	.8338943
state_15	.0353566	.4330996	0.08	0.935	-.8137077	.8844209
state_16	-.0882641	.4391458	-0.20	0.841	-.9491815	.7726532
state_17	.0229865	.4462893	0.05	0.959	-.8519352	.8979083
state_18	.1134403	.4460336	0.25	0.799	-.7609803	.9878609
state_19	-.1828887	.4524857	-0.40	0.686	-1.069958	.7041808
state_20	.005086	.4393396	0.01	0.991	-.8562112	.8663833
state_21	.2104414	.4385313	0.48	0.631	-.6492713	1.070154
state_22	.1884142	.4624072	0.41	0.684	-.7181053	1.094934
state_23	-.1565164	.433778	-0.36	0.718	-1.006911	.6938779
state_24	-.1572061	.4423172	-0.36	0.722	-1.024341	.7099287
state_25	-.1398459	.4386957	-0.32	0.750	-.999881	.7201892
state_26	.1893229	.4552749	0.42	0.678	-.7032143	1.08186
state_27	-.1820614	.479637	-0.38	0.704	-1.122359	.758236
state_28	-.0739984	.4350948	-0.17	0.865	-.9269743	.7789775
state_29	-.5851356	.6015479	-0.97	0.331	-1.76443	.5941589
state_30	-.1009495	.4675424	-0.22	0.829	-1.017536	.8156372
state_31	-.0845011	.4694222	-0.18	0.857	-1.004773	.8357708
state_32	.0581711	.4356912	0.13	0.894	-.7959738	.912316
state_33	-.177138	.4599256	-0.39	0.700	-1.078792	.7245164
state_34	-.095732	.4497081	-0.21	0.831	-.9773559	.7858919
state_35	.1282398	.4312848	0.30	0.766	-.7172668	.9737463
state_36	.0694005	.4336061	0.16	0.873	-.7806568	.9194579
state_37	.2267367	.4585931	0.49	0.621	-.6723057	1.125779
state_38	-.1327956	.4435003	-0.30	0.765	-1.00225	.7366585
state_39	.0341027	.4317258	0.08	0.937	-.8122685	.8804739
state_40	.4050129	.4928611	0.82	0.411	-.5612094	1.371235
state_41	-.0318962	.4446172	-0.07	0.943	-.90354	.8397476
state_42	.0990087	.4752477	0.21	0.835	-.8326838	1.030701

state_43	-.0339117	.4397639	-0.08	0.939	-.8960408	.8282175
state_44	-.0812478	.4306372	-0.19	0.850	-.9254849	.7629893
state_45	-.1688342	.4511007	-0.37	0.708	-1.053189	.7155202
state_46	-.0621671	.4368533	-0.14	0.887	-.9185903	.794256
state_47	.3000364	.5076999	0.59	0.555	-.6952759	1.295349
state_48	-.2214452	.435639	-0.51	0.611	-1.075488	.6325973
state_49	.0230673	.437251	0.05	0.958	-.8341357	.8802702
state_50	.0669568	.455053	0.15	0.883	-.8251456	.9590592
state_51	-.1747841	.5220006	-0.33	0.738	-1.198132	.848564
_cons	.4143258	.4368784	0.95	0.343	-.4421476	1.270799

```
26 .
27 . *Concern, Latent NSC 2
28 . mi estimate: regress z_dv_concern c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
>  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        965
                                                Average RVI       =     0.0112
                                                Largest FMI       =     0.0415
                                                Complete DF       =        946
DF adjustment:   Small sample                   DF:     min       =     850.29
                                                        avg       =     926.16
                                                        max       =     943.25
Model F test:       Equal FMI                   F(  18,  943.7)   =       2.70
Within VCE type:          OLS                   Prob > F          =     0.0002
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.045829	.0370731	-1.24	0.217	-.1185843	.0269262
z_polengage_scale	.0029868	.1936837	0.02	0.988	-.3771217	.3830952
c.z_latentnsc2#c.z_polengage_scale	-.0923728	.0479984	-1.92	0.055	-.1865689	.0018232
age	.045498	.1692745	0.27	0.788	-.2867003	.3776963
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1890517	.2052814	-0.92	0.357	-.5919134	.21381
1.male	-.1672555	.070208	-2.38	0.017	-.3050376	-.0294734
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.120961	.0869399	1.39	0.164	-.0496573	.2915793
race						
1	-.1674628	.1474239	-1.14	0.256	-.4567855	.1218599
2	.2456624	.1850288	1.33	0.185	-.117476	.6088007
3	.0587673	.1791187	0.33	0.743	-.2927629	.4102975
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0884646	.17277	0.51	0.609	-.2506052	.4275345
2	-.1509971	.2296966	-0.66	0.511	-.6018358	.2998417
3	-.1272016	.2079613	-0.61	0.541	-.5353463	.2809431
education	.3574676	.155128	2.30	0.021	.0530287	.6619065
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.1408716	.186989	0.75	0.451	-.2260966	.5078398
income	-.2664761	.131534	-2.03	0.043	-.5246143	-.008338
z_polengage_scale	0	(omitted)				

c.income#c.z_polengage_scale	-.0345328	.1466231	-0.24	0.814	-.3222848	.2532193
5.sample_num	-.0620829	.0804159	-0.77	0.440	-.2198981	.0957323
_cons	-.1848749	.1799615	-1.03	0.305	-.5380511	.1683013

```
29 . mi estimate: regress z_dv_concern c.z_latentnsc2##c.z_polint_scale c.age##c.z_polint_scale i.male##c.z_polint_scale i.
> ##c.z_polint_scale c.income##c.z_polint_scale i.sample_num

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =        896
                                              Average RVI       =     0.0110
                                              Largest FMI       =     0.0500
                                              Complete DF       =        877
DF adjustment:   Small sample                 DF:     min       =     766.09
                                                      avg       =     859.85
                                                      max       =     874.52
Model F test:       Equal FMI                 F(  18,  874.7)   =       2.87
Within VCE type:          OLS                 Prob > F          =     0.0001
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.0675916	.0360917	-1.87	0.061	-.138428	.0032449
z_polint_scale	.2342556	.2101685	1.11	0.265	-.1782477	.646759
c.z_latentnsc2#c.z_polint_scale	-.1087599	.046595	-2.33	0.020	-.2002111	-.0173087
age	-.0245849	.1664691	-0.15	0.883	-.3513106	.3021409
z_polint_scale	0	(omitted)				
c.age#c.z_polint_scale	-.1004997	.2109939	-0.48	0.634	-.5146139	.3136145
1.male	-.158981	.0724393	-2.19	0.028	-.3011563	-.0168057
z_polint_scale	0	(omitted)				
male#c.z_polint_scale						
1	.0887807	.0892107	1.00	0.320	-.0863118	.2638732
race						
1	-.213597	.1596372	-1.34	0.181	-.5269195	.0997255
2	.1833742	.1993134	0.92	0.358	-.2078323	.5745807
3	-.0508628	.1942669	-0.26	0.794	-.4321566	.330431
z_polint_scale	0	(omitted)				
race#c.z_polint_scale						
1	-.0625873	.1854784	-0.34	0.736	-.426637	.3014624
2	-.1520586	.2427629	-0.63	0.531	-.6286181	.3245009
3	-.1704886	.2152603	-0.79	0.429	-.5929995	.2520223
education	.3053013	.1624132	1.88	0.060	-.0134655	.6240681
z_polint_scale	0	(omitted)				
c.education#c.z_polint_scale	.0261201	.1873202	0.14	0.889	-.3415347	.3937749
income	-.271995	.1359094	-2.00	0.046	-.5387446	-.0052453
z_polint_scale	0	(omitted)				
c.income#c.z_polint_scale	-.0197124	.1495566	-0.13	0.895	-.3132482	.2738233
5.sample_num	-.0679926	.0822902	-0.83	0.409	-.2295025	.0935174
_cons	-.0312834	.1903421	-0.16	0.869	-.4048687	.342302

```
30 . mi estimate: regress z_dv_concern c.z_latentnsc2##c.z_polknow_scale c.age##c.z_polknow_scale i.male##c.z_polknow_scale
> tion##c.z_polknow_scale c.income##c.z_polknow_scale i.sample_num

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =        897
                                              Average RVI       =     0.0058
                                              Largest FMI       =     0.0144
                                              Complete DF       =        878
DF adjustment:   Small sample                 DF:     min       =     857.16
                                                      avg       =     868.87
                                                      max       =     874.87
Model F test:       Equal FMI                 F(  18,  875.9)   =       2.30
Within VCE type:          OLS                 Prob > F          =     0.0017
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.0782661	.0407854	-1.92	0.055	-.1583148	.0017825
z_polknow_scale	-.1745512	.1866007	-0.94	0.350	-.5407904	.191688
c.z_latentnsc2#c.z_polknow_scale	-.0288812	.0468034	-0.62	0.537	-.1207413	.062979
age	.1849006	.1812264	1.02	0.308	-.1707889	.54059
z_polknow_scale	0	(omitted)				
c.age#c.z_polknow_scale	-.1932939	.1922246	-1.01	0.315	-.5705695	.1839816
1.male	-.1454094	.0766154	-1.90	0.058	-.295781	.0049622
z_polknow_scale	0	(omitted)				
male#c.z_polknow_scale						
1	.0710294	.0842399	0.84	0.399	-.0943065	.2363654
race						
1	-.1612705	.1503948	-1.07	0.284	-.4564505	.1339094
2	.2613169	.1900179	1.38	0.169	-.1116378	.6342717
3	.1129929	.1827731	0.62	0.537	-.245739	.4717249
z_polknow_scale	0	(omitted)				
race#c.z_polknow_scale						
1	.1745458	.1586055	1.10	0.271	-.1367495	.485841
2	-.0905346	.2014048	-0.45	0.653	-.4858319	.3047626
3	-.0262576	.19434	-0.14	0.893	-.4076907	.3551755
education	.3902034	.1653494	2.36	0.019	.065672	.7147347
z_polknow_scale	0	(omitted)				
c.education#c.z_polknow_scale	.1319687	.1832703	0.72	0.472	-.2277364	.4916738
income	-.2869128	.1430566	-2.01	0.045	-.5676915	-.0061342
z_polknow_scale	0	(omitted)				
c.income#c.z_polknow_scale	.0111195	.1437478	0.08	0.938	-.2710192	.2932582
5.sample_num	-.0202902	.0840892	-0.24	0.809	-.1853308	.1447504
_cons	-.3084854	.1838634	-1.68	0.094	-.6693536	.0523829

```
31 .
32 . *Restrictions, Authoritarianism
33 . mi estimate: regress z_dv_restrictions c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_pole
   > scale c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =      6,870
                                              Average RVI       =     0.0044
                                              Largest FMI       =     0.0927
                                              Complete DF       =       6800
DF adjustment:   Small sample                 DF:     min       =   1,942.29
                                                      avg       =   6,579.18
                                                      max       =   6,797.96
Model F test:       Equal FMI                 F(  69, 6797.4)   =      15.39
Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1452909	.011767	-12.35	0.000	-.168358	-.1222238
z_polengage_scale	.1527916	.0520009	2.94	0.003	.050853	.2547302
c.z_auth_scale#c.z_polengage_scale	-.1271633	.0116161	-10.95	0.000	-.1499345	-.1043921
age	-.3368018	.0602837	-5.59	0.000	-.4549775	-.2186262
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.2429063	.0596376	-4.07	0.000	-.3598287	-.1259839
1.male	-.1265237	.0244029	-5.18	0.000	-.1743618	-.0786856
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.1244346	.0240245	-5.18	0.000	-.1715307	-.0773386
race						
1	-.0627554	.0464051	-1.35	0.176	-.153724	.0282131
2	.4443708	.0593011	7.49	0.000	.3281221	.5606195
3	.1173453	.0589651	1.99	0.047	.0017553	.2329353
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	-.0103028	.0436517	-0.24	0.813	-.0958739	.0752683
2	.1521327	.0547166	2.78	0.005	.0448709	.2593945
3	-.0035929	.0547645	-0.07	0.948	-.1109485	.1037626
education	.222088	.0466619	4.76	0.000	.1306142	.3135618
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0347358	.0461281	0.75	0.451	-.0556919	.1251636
income	.1078813	.0510345	2.11	0.035	.0077932	.2079694
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0788126	.0501228	-1.57	0.116	-.1771063	.0194811
sample_num						
4	-.2618038	.036557	-7.16	0.000	-.3334669	-.1901407
6	.0468147	.0353906	1.32	0.186	-.0225619	.1161914
state_1	.4072915	.3133021	1.30	0.194	-.2068787	1.021462
state_2	.0768143	.2596744	0.30	0.767	-.4322289	.5858574
state_3	.2696871	.2640055	1.02	0.307	-.2478462	.7872204
state_4	.0741794	.2508185	0.30	0.767	-.4175033	.5658622
state_5	.4319752	.2445196	1.77	0.077	-.0473598	.9113103
state_6	.2410412	.2548834	0.95	0.344	-.2586099	.7406924

state_7	.3056492	.2692725	1.14	0.256	-.2222092	.8335075
state_8	.8973029	.3112066	2.88	0.004	.2872405	1.507365
state_9	.3374615	.3057393	1.10	0.270	-.2618832	.9368063
state_10	.2210828	.2449053	0.90	0.367	-.2590081	.7011738
state_11	.2092814	.2496649	0.84	0.402	-.2801398	.6987027
state_12	.5337257	.2941762	1.81	0.070	-.0429517	1.110403
state_13	.3284876	.2651732	1.24	0.215	-.1913349	.8483102
state_14	.28373	.2789396	1.02	0.309	-.2630789	.8305389
state_15	.2720328	.2492047	1.09	0.275	-.2164864	.7605519
state_16	.1580365	.2544711	0.62	0.535	-.3408064	.6568794
state_17	.3520738	.2661909	1.32	0.186	-.1697437	.8738912
state_18	.3761117	.2612336	1.44	0.150	-.1359879	.8882113
state_19	-.121327	.2652138	-0.46	0.647	-.641229	.398575
state_20	.3748697	.2542578	1.47	0.140	-.1235552	.8732947
state_21	.4477444	.2548176	1.76	0.079	-.0517779	.9472666
state_22	.4132143	.284173	1.45	0.146	-.1438536	.9702823
state_23	.2169465	.2494538	0.87	0.385	-.2720611	.7059541
state_24	.3004684	.258954	1.16	0.246	-.2071624	.8080993
state_25	.0247084	.2532687	0.10	0.922	-.4717776	.5211944
state_26	-.0996606	.2784255	-0.36	0.720	-.6454617	.4461405
state_27	-.1254538	.2959003	-0.42	0.672	-.7055112	.4546035
state_28	.2004933	.2507482	0.80	0.424	-.2910517	.6920382
state_29	-.1326059	.3473913	-0.38	0.703	-.8136017	.5483898
state_30	.2004666	.27881	0.72	0.472	-.3460884	.7470215
state_31	.0233323	.2887769	0.08	0.936	-.5427609	.5894255
state_32	.1976629	.2510897	0.79	0.431	-.2945515	.6898773
state_33	-.1895692	.280941	-0.67	0.500	-.7403015	.3611632
state_34	.1435594	.2653047	0.54	0.588	-.3765208	.6636396
state_35	.5272921	.2456547	2.15	0.032	.0457319	1.008852
state_36	.2887025	.2489793	1.16	0.246	-.1993749	.7767799
state_37	.2588602	.2695181	0.96	0.337	-.2694796	.7872
state_38	.2164688	.2602985	0.83	0.406	-.2937978	.7267354
state_39	.1623558	.2463502	0.66	0.510	-.3205676	.6452793
state_40	.6408417	.3104673	2.06	0.039	.0322285	1.249455
state_41	.3056912	.2596173	1.18	0.239	-.2032399	.8146224
state_42	-.1962383	.3078353	-0.64	0.524	-.7996918	.4072153
state_43	.2020736	.2549374	0.79	0.428	-.2976835	.7018306
state_44	.211868	.2454274	0.86	0.388	-.2692465	.6929825
state_45	.0434388	.2701565	0.16	0.872	-.4861526	.5730302
state_46	.1980058	.2521727	0.79	0.432	-.2963317	.6923434
state_47	.3287067	.3134745	1.05	0.294	-.2858015	.9432148
state_48	.1921053	.2521689	0.76	0.446	-.3022246	.6864353
state_49	.1918321	.2536139	0.76	0.449	-.3053306	.6889947
state_50	.3394563	.2746562	1.24	0.217	-.1989558	.8778684
state_51	0	(omitted)				
_cons	-.1555637	.2500558	-0.62	0.534	-.6457515	.334624

```
34 . mi estimate: regress z_dv_restrictions c.z_auth_scale##c.z_polint_scale c.age##c.z_polint_scale i.male##c.z_polint_sca
   > ation##c.z_polint_scale c.income##c.z_polint_scale i.sample_num state_1-state_51

   Multiple-imputation estimates                 Imputations       =         25
   Linear regression                             Number of obs     =      6,868
                                                 Average RVI       =     0.0041
                                                 Largest FMI       =     0.0934
                                                 Complete DF       =       6798
   DF adjustment:   Small sample                 DF:     min       =   1,922.24
                                                         avg       =   6,591.37
                                                         max       =   6,795.84
   Model F test:       Equal FMI                 F(  69, 6795.5)   =      14.61
   Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1437163	.011717	-12.27	0.000	-.1666853	-.1207474
z_polint_scale	.1424251	.0509436	2.80	0.005	.0425594	.2422908
c.z_auth_scale#c.z_polint_scale	-.0960328	.011422	-8.41	0.000	-.1184235	-.073642
age	-.3515043	.0600324	-5.86	0.000	-.4691874	-.2338213
z_polint_scale	0	(omitted)				
c.age#c.z_polint_scale	-.3237423	.0570649	-5.67	0.000	-.4356148	-.2118699
1.male	-.159325	.0241659	-6.59	0.000	-.2066983	-.1119517
z_polint_scale	0	(omitted)				
male#c.z_polint_scale						
1	-.0868244	.0235672	-3.68	0.000	-.1330241	-.0406248
race						
1	-.0849355	.0466611	-1.82	0.069	-.1764058	.0065348
2	.4115192	.0588917	6.99	0.000	.2960731	.5269654
3	.1108248	.0587718	1.89	0.059	-.0043864	.226036
z_polint_scale	0	(omitted)				
race#c.z_polint_scale						
1	.0428236	.0425254	1.01	0.314	-.0405395	.1261867
2	.1762894	.0541075	3.26	0.001	.0702217	.282357
3	.0244956	.0534157	0.46	0.647	-.0802159	.1292071
education	.1937802	.046505	4.17	0.000	.1026137	.2849466
z_polint_scale	0	(omitted)				
c.education#c.z_polint_scale	.06591	.0454123	1.45	0.147	-.0231145	.1549344
income	.0819476	.0513166	1.60	0.110	-.0186945	.1825896
z_polint_scale	0	(omitted)				
c.income#c.z_polint_scale	.0035763	.0494232	0.07	0.942	-.0933478	.1005003
sample_num						
4	-.2552901	.0366871	-6.96	0.000	-.3272083	-.1833719
6	.0117959	.0355889	0.33	0.740	-.0579695	.0815613
state_1	.3887882	.3144472	1.24	0.216	-.2276268	1.005203
state_2	.0676971	.2605759	0.26	0.795	-.4431133	.5785074
state_3	.2738693	.2649604	1.03	0.301	-.245536	.7932746
state_4	.0489724	.2517355	0.19	0.846	-.4445081	.5424529
state_5	.4171944	.2454331	1.70	0.089	-.0639313	.89832
state_6	.2332078	.2558096	0.91	0.362	-.2682591	.7346747
state_7	.276221	.270284	1.02	0.307	-.2536203	.8060623
state_8	.8808225	.3123276	2.82	0.005	.2685625	1.493082
state_9	.323797	.3069072	1.06	0.291	-.2778373	.9254313
state_10	.2088173	.2458083	0.85	0.396	-.2730439	.6906786
state_11	.1837308	.2505781	0.73	0.463	-.3074808	.6749424
state_12	.5186403	.2952485	1.76	0.079	-.0601392	1.09742
state_13	.349349	.2661683	1.31	0.189	-.1724243	.8711222
state_14	.258253	.2800004	0.92	0.356	-.2906354	.8071415
state_15	.2592624	.2501024	1.04	0.300	-.2310166	.7495414
state_16	.1549291	.2554406	0.61	0.544	-.3458144	.6556727
state_17	.3528298	.2671985	1.32	0.187	-.1709629	.8766226
state_18	.3637454	.2621798	1.39	0.165	-.1502091	.8777
state_19	-.1176536	.2661857	-0.44	0.659	-.6394608	.4041536
state_20	.3687874	.2551981	1.45	0.148	-.1314807	.8690555
state_21	.4482099	.2557447	1.75	0.080	-.0531298	.9495496
state_22	.4166167	.2852616	1.46	0.144	-.1425854	.9758188

state_23	.2110007	.2503606	0.84	0.399	-.2797845	.7017859
state_24	.2865517	.2598815	1.10	0.270	-.2228975	.7960009
state_25	.0197011	.254217	0.08	0.938	-.4786439	.518046
state_26	-.086757	.2794703	-0.31	0.756	-.6346064	.4610924
state_27	-.1579099	.2968955	-0.53	0.595	-.7399181	.4240982
state_28	.1784552	.2517005	0.71	0.478	-.3149566	.6718671
state_29	-.0929395	.3486751	-0.27	0.790	-.7764518	.5905729
state_30	.1804024	.2797781	0.64	0.519	-.3680502	.728855
state_31	.0163609	.289802	0.06	0.955	-.5517417	.5844636
state_32	.1870044	.2520187	0.74	0.458	-.3070312	.68104
state_33	-.2281594	.2818922	-0.81	0.418	-.7807564	.3244376
state_34	.1068346	.2662685	0.40	0.688	-.415135	.6288042
state_35	.5166526	.2465503	2.10	0.036	.0333369	.9999684
state_36	.2766734	.2498934	1.11	0.268	-.2131958	.7665427
state_37	.2687144	.2705162	0.99	0.321	-.2615821	.799011
state_38	.2141555	.2612492	0.82	0.412	-.2979747	.7262857
state_39	.1528734	.2472751	0.62	0.536	-.3318632	.6376099
state_40	.5814913	.3116682	1.87	0.062	-.029476	1.192459
state_41	.2971411	.2605732	1.14	0.254	-.213664	.8079462
state_42	-.1547826	.3090198	-0.50	0.616	-.7605582	.4509929
state_43	.2047424	.2559088	0.80	0.424	-.2969189	.7064038
state_44	.1967262	.2463412	0.80	0.425	-.2861798	.6796322
state_45	.0395731	.2711588	0.15	0.884	-.4919832	.5711293
state_46	.2015209	.2530999	0.80	0.426	-.2946341	.697676
state_47	.3179224	.3145251	1.01	0.312	-.2986452	.93449
state_48	.1683596	.2530681	0.67	0.506	-.3277332	.6644524
state_49	.1753368	.2545435	0.69	0.491	-.323648	.6743217
state_50	.3558136	.2756937	1.29	0.197	-.1846323	.8962596
state_51	0	(omitted)				
_cons	-.0719879	.2506519	-0.29	0.774	-.5633441	.4193683

35 . mi estimate: regress z_dv_restrictions c.z_auth_scale##c.z_polknow_scale c.age##c.z_polknow_scale i.male##c.z_polknow_
> education##c.z_polknow_scale c.income##c.z_polknow_scale i.sample_num state_1-state_51

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 6,523
Average RVI = 0.0023
Largest FMI = 0.0826
Complete DF = 6453
DF adjustment: **Small sample** — DF: min = 2,224.16
avg = 6,342.11
max = 6,450.96
Model F test: **Equal FMI** — F(69, 6450.9) = 14.54
Within VCE type: **OLS** — Prob > F = 0.0000

z_dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.1584596	.012009	-13.20	0.000	-.1820012	-.1349181
z_polknow_scale	.1172205	.0536558	2.18	0.029	.0120373	.2224038
c.z_auth_scale#c.z_polknow_scale	-.1183486	.0118766	-9.96	0.000	-.1416306	-.0950665
age	-.2422834	.0624196	-3.88	0.000	-.364647	-.1199198
z_polknow_scale	0	(omitted)				
c.age#c.z_polknow_scale	-.1624025	.0605562	-2.68	0.007	-.2811133	-.0436917
1.male	-.1078499	.0248187	-4.35	0.000	-.1565028	-.059197
z_polknow_scale	0	(omitted)				
male#c.z_polknow_scale						
1	-.1210451	.0246555	-4.91	0.000	-.1693782	-.072712
race						
1	-.0567524	.0474844	-1.20	0.232	-.1498377	.0363328

2	.4530638	.0623229	7.27	0.000	.3308902	.5752374
3	.1145975	.0607045	1.89	0.059	-.0044034	.2335983
z_polknow_scale	0	(omitted)				
race#c.z_polknow_scale						
1	-.0572236	.0447051	-1.28	0.201	-.1448605	.0304132
2	.081517	.0561443	1.45	0.147	-.0285444	.1915784
3	-.0224005	.0562065	-0.40	0.690	-.132584	.087783
education	.2346522	.0475666	4.93	0.000	.1414043	.3279
z_polknow_scale	0	(omitted)				
c.education#c.z_polknow_scale	.0276709	.0469382	0.59	0.556	-.0643443	.1196861
income	.124912	.0518369	2.41	0.016	.0232582	.2265658
z_polknow_scale	0	(omitted)				
c.income#c.z_polknow_scale	-.0892701	.0487124	-1.83	0.067	-.1847715	.0062312
sample_num						
4	-.2739212	.0367633	-7.45	0.000	-.3459895	-.2018528
6	.0626568	.0352498	1.78	0.076	-.0064445	.131758
state_1	.437484	.3234289	1.35	0.176	-.196544	1.071512
state_2	.1072665	.2612299	0.41	0.681	-.4048307	.6193637
state_3	.2528368	.2653897	0.95	0.341	-.2674151	.7730886
state_4	.0672343	.2516309	0.27	0.789	-.4260458	.5605143
state_5	.4448325	.2448617	1.82	0.069	-.0351776	.9248426
state_6	.267558	.2554365	1.05	0.295	-.2331824	.7682984
state_7	.3111431	.2707315	1.15	0.250	-.2195805	.8418667
state_8	.9230681	.3114958	2.96	0.003	.3124329	1.533703
state_9	.3697519	.305841	1.21	0.227	-.229798	.9693017
state_10	.2311711	.2451867	0.94	0.346	-.2494762	.7118184
state_11	.2459683	.2503487	0.98	0.326	-.2447982	.7367348
state_12	.6300535	.2960215	2.13	0.033	.0497531	1.210354
state_13	.3238058	.2656518	1.22	0.223	-.1969598	.8445715
state_14	.3363758	.2815498	1.19	0.232	-.2155552	.8883068
state_15	.2510146	.2496562	1.01	0.315	-.2383943	.7404235
state_16	.156963	.2557158	0.61	0.539	-.3443249	.6582509
state_17	.3557752	.2670197	1.33	0.183	-.1676721	.8792225
state_18	.4029952	.2623098	1.54	0.125	-.1112191	.9172095
state_19	-.1215783	.2657356	-0.46	0.647	-.6425082	.3993516
state_20	.4173237	.2550919	1.64	0.102	-.082741	.9173884
state_21	.499807	.2557162	1.95	0.051	-.0014817	1.001096
state_22	.4273611	.2843394	1.50	0.133	-.1300384	.9847606
state_23	.2499329	.2502237	1.00	0.318	-.2405885	.7404544
state_24	.3375003	.2604101	1.30	0.195	-.1729899	.8479904
state_25	.0366329	.2541611	0.14	0.885	-.4616071	.534873
state_26	-.0601195	.2827206	-0.21	0.832	-.6143456	.4941067
state_27	-.115034	.2960784	-0.39	0.698	-.6954459	.4653779
state_28	.2323795	.2514553	0.92	0.355	-.2605564	.7253153
state_29	-.1341993	.3476604	-0.39	0.700	-.8157292	.5473305
state_30	.17591	.2815121	0.62	0.532	-.3759472	.7277672
state_31	-.0002721	.2902625	-0.00	0.999	-.5692829	.5687388
state_32	.2133604	.2517579	0.85	0.397	-.2801685	.7068894
state_33	-.1411173	.2828965	-0.50	0.618	-.6956883	.4134537
state_34	.1414471	.2661065	0.53	0.595	-.38021	.6631041
state_35	.5446122	.2460111	2.21	0.027	.0623488	1.026876
state_36	.322194	.2493795	1.29	0.196	-.1666726	.8110607
state_37	.326428	.2720657	1.20	0.230	-.206911	.8597669
state_38	.2640007	.2610609	1.01	0.312	-.2477653	.7757667
state_39	.1718197	.2467311	0.70	0.486	-.311855	.6554944
state_40	.7234763	.3134637	2.31	0.021	.1089833	1.337969
state_41	.3017351	.2605214	1.16	0.247	-.2089734	.8124435
state_42	-.2226079	.3080086	-0.72	0.470	-.826407	.3811913
state_43	.1687696	.2562896	0.66	0.510	-.333643	.6711822

state_44	.2325244	.2458021	0.95	0.344	-.2493292	.714378
state_45	.0214508	.2727508	0.08	0.937	-.5132312	.5561328
state_46	.2137125	.2530041	0.84	0.398	-.2822596	.7096846
state_47	.4116911	.3199239	1.29	0.198	-.2154658	1.038848
state_48	.209355	.2525825	0.83	0.407	-.2857905	.7045004
state_49	.2406045	.2541795	0.95	0.344	-.2576716	.7388807
state_50	.354167	.2754664	1.29	0.199	-.1858385	.8941726
state_51	0	(omitted)				
_cons	-.2482542	.2503936	-0.99	0.322	-.7391088	.2426003

```
36 .
37 . *Behaviors, Authoritarianism
38 . mi estimate: regress z_dv_behavior c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
   > e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num state_1-state_51

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =      4,029
                                                Average RVI       =     0.0040
                                                Largest FMI       =     0.0580
                                                Complete DF       =       3957
DF adjustment:   Small sample                   DF:     min       =   2,456.86
                                                        avg       =   3,869.56
                                                        max       =   3,950.96
Model F test:       Equal FMI                   F(  71, 3954.8)   =      15.08
Within VCE type:          OLS                   Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.0611614	.0169203	-3.61	0.000	-.0943348	-.027988
z_polengage_scale	.3436187	.0673692	5.10	0.000	.2115366	.4757007
c.z_auth_scale#c.z_polengage_scale	-.0387107	.0166962	-2.32	0.020	-.0714447	-.0059768
age	.0865129	.0732158	1.18	0.237	-.0570351	.2300609
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.4430828	.0733348	-6.04	0.000	-.5868827	-.2992829
1.male	-.1048385	.0316475	-3.31	0.001	-.166886	-.042791
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.0471364	.031825	-1.48	0.139	-.1095352	.0152624
race						
1	-.0653641	.0618411	-1.06	0.291	-.1866079	.0558797
2	.2345092	.0792877	2.96	0.003	.0790604	.389958
3	.0668377	.0747687	0.89	0.371	-.0797515	.2134268
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0248582	.0603894	0.41	0.681	-.0935394	.1432559
2	-.0406227	.0745013	-0.55	0.586	-.1866874	.1054419
3	-.1223424	.0709286	-1.72	0.085	-.2614027	.016718
education	.2054511	.0640001	3.21	0.001	.0799746	.3309275
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	-.0298052	.0626569	-0.48	0.634	-.1526513	.093041
income	.2413437	.057582	4.19	0.000	.1284487	.3542386
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0600431	.0581653	-1.03	0.302	-.1741011	.0540149

sample_num						
3	.048299	.053668	0.90	0.368	-.0569207	.1535187
4	-.6034641	.0490528	-12.30	0.000	-.6996352	-.5072929
5	-.0011198	.0619062	-0.02	0.986	-.122491	.1202514
state_1	.474406	.5735015	0.83	0.408	-.6499819	1.598794
state_2	.0458776	.4733168	0.10	0.923	-.8820924	.9738475
state_3	.4027713	.4801739	0.84	0.402	-.5386425	1.344185
state_4	.2489446	.4668835	0.53	0.594	-.6664126	1.164302
state_5	.5385585	.4608952	1.17	0.243	-.3650582	1.442175
state_6	.0792465	.471251	0.17	0.866	-.8446734	1.003166
state_7	.4964547	.4836888	1.03	0.305	-.4518502	1.44476
state_8	.6514098	.5039688	1.29	0.196	-.3366553	1.639475
state_9	.4800624	.5192845	0.92	0.355	-.5380299	1.498155
state_10	.3079503	.4615841	0.67	0.505	-.5970171	1.212918
state_11	.3163131	.4647806	0.68	0.496	-.5949211	1.227547
state_12	1.021017	.506014	2.02	0.044	.0289417	2.013091
state_13	-.0496464	.4826736	-0.10	0.918	-.9959608	.8966681
state_14	.5732536	.5119995	1.12	0.263	-.4305559	1.577063
state_15	.2425983	.4650975	0.52	0.602	-.6692574	1.154454
state_16	.1380123	.470052	0.29	0.769	-.7835568	1.059582
state_17	.0905594	.4792515	0.19	0.850	-.849046	1.030165
state_18	.3802245	.4738101	0.80	0.422	-.5487126	1.309162
state_19	.2137155	.4758387	0.45	0.653	-.7191989	1.14663
state_20	.4284017	.4710662	0.91	0.363	-.495156	1.351959
state_21	.365243	.4706688	0.78	0.438	-.5575353	1.288021
state_22	.5488929	.5125838	1.07	0.284	-.4560624	1.553848
state_23	.2757817	.4660124	0.59	0.554	-.6378677	1.189431
state_24	.2276042	.4754722	0.48	0.632	-.7045912	1.159799
state_25	.1323541	.4695207	0.28	0.778	-.7881735	1.052882
state_26	.195907	.4893547	0.40	0.689	-.7635064	1.15532
state_27	-.0867293	.5206117	-0.17	0.868	-1.107424	.9339656
state_28	.28833	.4655705	0.62	0.536	-.6244529	1.201113
state_29	-.3469409	.5352876	-0.65	0.517	-1.396408	.7025264
state_30	.2458721	.4971128	0.49	0.621	-.7287513	1.220496
state_31	-.2259033	.51238	-0.44	0.659	-1.230459	.7786525
state_32	.426272	.4664933	0.91	0.361	-.4883201	1.340864
state_33	.308412	.5170749	0.60	0.551	-.7053517	1.322176
state_34	.3009821	.4770707	0.63	0.528	-.6343477	1.236312
state_35	.5295048	.4612549	1.15	0.251	-.3748173	1.433827
state_36	.2509619	.4647157	0.54	0.589	-.6601453	1.162069
state_37	.1480037	.4770375	0.31	0.756	-.7872612	1.083268
state_38	.4551638	.4752014	0.96	0.338	-.4765012	1.386829
state_39	.2676028	.4633561	0.58	0.564	-.6408389	1.176044
state_40	.2122995	.5197614	0.41	0.683	-.8067277	1.231327
state_41	.3040367	.4723436	0.64	0.520	-.6220253	1.230099
state_42	-.9690191	.6146758	-1.58	0.115	-2.174132	.2360937
state_43	.0773172	.4716075	0.16	0.870	-.8473019	1.001936
state_44	.2996923	.4613846	0.65	0.516	-.604884	1.204269
state_45	.1552291	.4878598	0.32	0.750	-.8012532	1.111711
state_46	.2912784	.4682874	0.62	0.534	-.626831	1.209388
state_47	.2114707	.5423908	0.39	0.697	-.8519229	1.274864
state_48	.3798505	.4699061	0.81	0.419	-.5414324	1.301133
state_49	.2858859	.4737588	0.60	0.546	-.6429508	1.214723
state_50	.3662946	.4881529	0.75	0.453	-.5907624	1.323352
state_51	-.1152303	.5419889	-0.21	0.832	-1.177836	.9473753
_cons	-.2147736	.4659779	-0.46	0.645	-1.128355	.698808

```
39 . mi estimate: regress z_dv_behavior c.z_auth_scale##c.z_polint_scale c.age##c.z_polint_scale i.male##c.z_polint_scale i
> n##c.z_polint_scale c.income##c.z_polint_scale i.sample_num state_1-state_51

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =      3,960
                                              Average RVI       =     0.0034
                                              Largest FMI       =     0.0418
                                              Complete DF       =       3888
DF adjustment:   Small sample                 DF:     min       =   2,935.98
                                                      avg       =   3,748.00
                                                      max       =   3,883.19
Model F test:       Equal FMI                 F(  71, 3885.9)   =      14.91
Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.0582823	.0172108	-3.39	0.001	-.0920253	-.0245393
z_polint_scale	.2137695	.0640725	3.34	0.001	.0881501	.3393889
c.z_auth_scale#c.z_polint_scale	-.016026	.015695	-1.02	0.307	-.0467973	.0147452
age	.1280272	.0717884	1.78	0.075	-.0127224	.2687768
z_polint_scale	0	(omitted)				
c.age#c.z_polint_scale	-.4057183	.0677881	-5.99	0.000	-.5386337	-.272803
1.male	-.1073044	.0324553	-3.31	0.001	-.170936	-.0436728
z_polint_scale	0	(omitted)				
male#c.z_polint_scale						
1	-.0146934	.0307606	-0.48	0.633	-.0750048	.0456181
race						
1	-.052919	.0640557	-0.83	0.409	-.1785052	.0726671
2	.2428779	.079399	3.06	0.002	.08721	.3985457
3	.0930566	.0758472	1.23	0.220	-.0556478	.241761
z_polint_scale	0	(omitted)				
race#c.z_polint_scale						
1	.0792928	.0564145	1.41	0.160	-.0313123	.1898979
2	.0509084	.0716711	0.71	0.478	-.0896083	.1914252
3	-.0372183	.0667903	-0.56	0.577	-.168166	.0937295
education	.2372315	.0653293	3.63	0.000	.1091485	.3653145
z_polint_scale	0	(omitted)				
c.education#c.z_polint_scale	.0247996	.0593175	0.42	0.676	-.091498	.1410971
income	.2370438	.0591697	4.01	0.000	.1210357	.353052
z_polint_scale	0	(omitted)				
c.income#c.z_polint_scale	-.0003132	.0561042	-0.01	0.996	-.1103207	.1096944
sample_num						
3	.0281028	.0551019	0.51	0.610	-.0799286	.1361341
4	-.6317348	.0504557	-12.52	0.000	-.730657	-.5328126
5	.0230652	.0638269	0.36	0.718	-.1020722	.1482027
state_1	.7687235	.7362924	1.04	0.297	-.6748442	2.212291
state_2	.3466929	.6608379	0.52	0.600	-.9489437	1.642329
state_3	.7087721	.6656619	1.06	0.287	-.596322	2.013866
state_4	.5456188	.6559605	0.83	0.406	-.7404558	1.831693
state_5	.8274626	.6518208	1.27	0.204	-.4504958	2.105421
state_6	.3702025	.6590344	0.56	0.574	-.921898	1.662303
state_7	.8077995	.6685377	1.21	0.227	-.5029328	2.118532

state_8	.9200791	.6831953	1.35	0.178	-.4193898	2.259548
state_9	.7852237	.6943098	1.13	0.258	-.5760356	2.146483
state_10	.5896477	.6522521	0.90	0.366	-.6891563	1.868452
state_11	.5952087	.654723	0.91	0.363	-.6884391	1.878857
state_12	1.347107	.6844796	1.97	0.049	.0051207	2.689094
state_13	.2356739	.6673827	0.35	0.724	-1.072794	1.544142
state_14	.8756767	.691463	1.27	0.205	-.4800011	2.231354
state_15	.5305195	.6548797	0.81	0.418	-.7534364	1.814475
state_16	.4619405	.6585407	0.70	0.483	-.8291923	1.753073
state_17	.4869586	.6650636	0.73	0.464	-.8169625	1.79088
state_18	.678558	.6608312	1.03	0.305	-.6170653	1.974181
state_19	.5325251	.6627792	0.80	0.422	-.7669176	1.831968
state_20	.7381882	.6589985	1.12	0.263	-.5538422	2.030219
state_21	.6697567	.6589297	1.02	0.309	-.622139	1.961652
state_22	.8568637	.6892065	1.24	0.214	-.4943906	2.208118
state_23	.5973081	.6557182	0.91	0.362	-.6882912	1.882907
state_24	.5230075	.6624257	0.79	0.430	-.7757418	1.821757
state_25	.4660781	.6580308	0.71	0.479	-.824055	1.756211
state_26	.4456714	.6743468	0.66	0.509	-.8764508	1.767794
state_27	.2209084	.695029	0.32	0.751	-1.141761	1.583577
state_28	.5791361	.6552477	0.88	0.377	-.7055407	1.863813
state_29	-.0122369	.7062371	-0.02	0.986	-1.39688	1.372406
state_30	.5557781	.677791	0.82	0.412	-.7730957	1.884652
state_31	.0903618	.6891811	0.13	0.896	-1.260843	1.441566
state_32	.7120669	.6559209	1.09	0.278	-.5739299	1.998064
state_33	.609786	.6922117	0.88	0.378	-.7473636	1.966936
state_34	.5847318	.6633517	0.88	0.378	-.7158331	1.885297
state_35	.8233436	.6520435	1.26	0.207	-.4550515	2.101739
state_36	.5475412	.6545902	0.84	0.403	-.7358469	1.830929
state_37	.4498515	.6632931	0.68	0.498	-.8505986	1.750302
state_38	.7527821	.661796	1.14	0.255	-.544733	2.050297
state_39	.5954467	.6535556	0.91	0.362	-.685913	1.876806
state_40	.4855184	.6945453	0.70	0.485	-.8762025	1.847239
state_41	.5854341	.6601962	0.89	0.375	-.7089445	1.879813
state_42	-.5041153	.7688279	-0.66	0.512	-2.01147	1.003239
state_43	.3834255	.6593992	0.58	0.561	-.9093907	1.676242
state_44	.5966375	.6521267	0.91	0.360	-.6819205	1.875195
state_45	.4478268	.6709597	0.67	0.505	-.8676538	1.763307
state_46	.5980079	.6574154	0.91	0.363	-.6909183	1.886934
state_47	.508624	.7116954	0.71	0.475	-.8867203	1.903968
state_48	.7023594	.6582438	1.07	0.286	-.5881913	1.99291
state_49	.6317239	.6613309	0.96	0.340	-.6648793	1.928327
state_50	.6611055	.6717804	0.98	0.325	-.6559835	1.978195
state_51	.1805604	.7114233	0.25	0.800	-1.21425	1.575371
_cons	-.5500738	.65494	-0.84	0.401	-1.834146	.7339989

```
40 . mi estimate: regress z_dv_behavior c.z_auth_scale##c.z_polknow_scale c.age##c.z_polknow_scale i.male##c.z_polknow_scal
> ation##c.z_polknow_scale c.income##c.z_polknow_scale i.sample_num state_1-state_51

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =      3,944
                                              Average RVI       =     0.0024
                                              Largest FMI       =     0.0305
                                              Complete DF       =       3872
DF adjustment:   Small sample                 DF:     min       =   3,279.82
                                                      avg       =   3,824.34
                                                      max       =   3,865.49
Model F test:       Equal FMI                 F(  71, 3869.9)   =      14.27
Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.06601	.0169956	-3.88	0.000	-.0993313	-.0326888
z_polknow_scale	.3333077	.0635125	5.25	0.000	.2087863	.4578291
c.z_auth_scale#c.z_polknow_scale	-.0467417	.0164473	-2.84	0.005	-.0789879	-.0144956
age	.2417623	.0736796	3.28	0.001	.097306	.3862185
z_polknow_scale	0	(omitted)				
c.age#c.z_polknow_scale	-.3362827	.0728767	-4.61	0.000	-.4791712	-.1933942
1.male	-.0831471	.0313813	-2.65	0.008	-.144673	-.0216212
z_polknow_scale	0	(omitted)				
male#c.z_polknow_scale						
1	-.0624322	.0303672	-2.06	0.040	-.1219703	-.0028941
race						
1	-.0746237	.059793	-1.25	0.212	-.1918528	.0426053
2	.2243985	.0782587	2.87	0.004	.0709661	.377831
3	.073858	.0727735	1.01	0.310	-.0688204	.2165364
z_polknow_scale	0	(omitted)				
race#c.z_polknow_scale						
1	-.0283249	.0559814	-0.51	0.613	-.138081	.0814312
2	-.1142936	.0688497	-1.66	0.097	-.2492788	.0206917
3	-.1688735	.0665234	-2.54	0.011	-.2992979	-.0384491
education	.2344119	.0636055	3.69	0.000	.1097081	.3591157
z_polknow_scale	0	(omitted)				
c.education#c.z_polknow_scale	-.0606396	.0611771	-0.99	0.322	-.1805834	.0593042
income	.2668596	.0575923	4.63	0.000	.1539433	.3797759
z_polknow_scale	0	(omitted)				
c.income#c.z_polknow_scale	-.1049601	.0543483	-1.93	0.054	-.2115188	.0015986
sample_num						
3	.0489159	.0552798	0.88	0.376	-.0594645	.1572963
4	-.6117853	.0507168	-12.06	0.000	-.7112196	-.512351
5	.0120833	.0644011	0.19	0.851	-.11418	.1383467
state_1	.4196986	.5762944	0.73	0.466	-.7101725	1.54957
state_2	-.0222406	.4755897	-0.05	0.963	-.9546726	.9101914
state_3	.3498923	.4825594	0.73	0.468	-.5962045	1.295989
state_4	.2094331	.4694211	0.45	0.656	-.7109052	1.129771
state_5	.5120626	.4632636	1.11	0.269	-.3962034	1.420329
state_6	.0397967	.4736064	0.08	0.933	-.8887472	.9683405
state_7	.4903781	.4859888	1.01	0.313	-.4624423	1.443198
state_8	.650709	.5065793	1.28	0.199	-.3424807	1.643899
state_9	.4304662	.5216598	0.83	0.409	-.5922897	1.453222
state_10	.2699354	.464013	0.58	0.561	-.6397998	1.179671
state_11	.2913773	.4674709	0.62	0.533	-.6251375	1.207892
state_12	.9452294	.5083811	1.86	0.063	-.0514926	1.941952
state_13	-.1210368	.4857536	-0.25	0.803	-1.073396	.8313223
state_14	.5745776	.5145468	1.12	0.264	-.4342328	1.583388
state_15	.1787105	.4677022	0.38	0.702	-.7382577	1.095679
state_16	.0989153	.4729929	0.21	0.834	-.8284256	1.026256
state_17	.047634	.4816786	0.10	0.921	-.896736	.992004
state_18	.3693074	.4769103	0.77	0.439	-.5657138	1.304329
state_19	.1396399	.4781858	0.29	0.770	-.7978821	1.077162
state_20	.3864238	.4736027	0.82	0.415	-.5421129	1.31496
state_21	.2958471	.4734305	0.62	0.532	-.6323519	1.224046

state_22	.5156271	.5150844	1.00	0.317	-.4942374	1.525492
state_23	.2444336	.4685865	0.52	0.602	-.6742684	1.163136
state_24	.1738921	.4778171	0.36	0.716	-.7629068	1.110691
state_25	.0855232	.4720336	0.18	0.856	-.839937	1.010983
state_26	.1687123	.4916163	0.34	0.731	-.7951415	1.132566
state_27	-.1527212	.5230857	-0.29	0.770	-1.178273	.8728304
state_28	.2735744	.468154	0.58	0.559	-.6442797	1.191428
state_29	-.4365953	.538007	-0.81	0.417	-1.491401	.6182106
state_30	.2071838	.4996149	0.41	0.678	-.7723515	1.186719
state_31	-.2986203	.5183991	-0.58	0.565	-1.314983	.7177426
state_32	.4047994	.4689661	0.86	0.388	-.5146468	1.324246
state_33	.2919624	.5194091	0.56	0.574	-.7263833	1.310308
state_34	.2681972	.4798168	0.56	0.576	-.6725225	1.208917
state_35	.4849516	.4635928	1.05	0.296	-.4239597	1.393863
state_36	.2062112	.4671796	0.44	0.659	-.7097325	1.122155
state_37	.0628838	.4802089	0.13	0.896	-.8786047	1.004372
state_38	.4146308	.4775791	0.87	0.385	-.5217019	1.350963
state_39	.2236483	.4656738	0.48	0.631	-.6893431	1.13664
state_40	.1908425	.5223085	0.37	0.715	-.8331851	1.21487
state_41	.2623409	.4749673	0.55	0.581	-.6688711	1.193553
state_42	-1.149651	.6172225	-1.86	0.063	-2.359765	.0604621
state_43	.0801519	.47412	0.17	0.866	-.8493989	1.009703
state_44	.2686886	.4637309	0.58	0.562	-.6404935	1.177871
state_45	.1046276	.4912851	0.21	0.831	-.8585764	1.067832
state_46	.2415218	.4708583	0.51	0.608	-.6816341	1.164678
state_47	.1637623	.5451822	0.30	0.764	-.9051112	1.232636
state_48	.3470405	.4724266	0.73	0.463	-.57919	1.273271
state_49	.2093995	.4767376	0.44	0.661	-.7252832	1.144082
state_50	.2798745	.4915515	0.57	0.569	-.6838518	1.243601
state_51	-.2035588	.5446342	-0.37	0.709	-1.271358	.8642402
_cons	-.2892023	.4680011	-0.62	0.537	-1.206756	.6283518

```
41 .
42 . *Behaviors, Latent NSC 2
43 . mi estimate: regress z_dv_behavior c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
   > e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num

Multiple-imputation estimates                   Imputations       =         25
Linear regression                               Number of obs     =        970
                                                Average RVI       =     0.0100
                                                Largest FMI       =     0.0224
                                                Complete DF       =        951
DF adjustment:   Small sample                   DF:     min       =     910.40
                                                        avg       =     933.13
                                                        max       =     948.29
Model F test:       Equal FMI                   F(  18,  948.7)   =       4.11
Within VCE type:          OLS                   Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.0792841	.0294006	-2.70	0.007	-.1369819	-.0215864
z_polengage_scale	.1522316	.1624136	0.94	0.349	-.1665041	.4709673
c.z_latentnsc2#c.z_polengage_scale	-.0476964	.0382142	-1.25	0.212	-.1226906	.0272977
age	-.124416	.1343222	-0.93	0.355	-.3880218	.1391899
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1727281	.1631452	-1.06	0.290	-.4928966	.1474404
1.male	-.2315128	.0565879	-4.09	0.000	-.3425658	-.1204597
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.0172722	.0704937	0.25	0.806	-.1210699	.1556143

race						
1	-.2025207	.1174573	-1.72	0.085	-.4330358	.0279944
2	-.0287809	.1472701	-0.20	0.845	-.3178022	.2602403
3	-.1124678	.14412	-0.78	0.435	-.3953133	.1703776
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	-.0058489	.1423268	-0.04	0.967	-.2851647	.2734668
2	.1137387	.18537	0.61	0.540	-.2500516	.4775289
3	.0824457	.1747175	0.47	0.637	-.2604477	.4253391
education	.2247866	.1271918	1.77	0.078	-.0248285	.4744017
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0546148	.1534042	0.36	0.722	-.2464409	.3556705
income	.3569805	.1071197	3.33	0.001	.1467532	.5672078
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0456332	.1196956	-0.38	0.703	-.2805446	.1892783
5.sample_num	-.0954953	.0647909	-1.47	0.141	-.2226464	.0316557
_cons	.3608952	.1436351	2.51	0.012	.0790087	.6427817

```
44 . mi estimate: regress z_dv_behavior c.z_latentnsc2##c.z_polint_scale c.age##c.z_polint_scale i.male##c.z_polint_scale i
> n##c.z_polint_scale c.income##c.z_polint_scale i.sample_num

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =        902
                                              Average RVI       =     0.0104
                                              Largest FMI       =     0.0218
                                              Complete DF       =        883
DF adjustment:   Small sample                 DF:     min       =     847.46
                                                      avg       =     866.37
                                                      max       =     880.29
Model F test:       Equal FMI                 F(  18,  880.7)   =       4.37
Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.092377	.028687	-3.22	0.001	-.14868	-.036074
z_polint_scale	.2462301	.1596754	1.54	0.123	-.067165	.5596251
c.z_latentnsc2#c.z_polint_scale	-.0453185	.0368367	-1.23	0.219	-.1176164	.0269795
age	-.1033677	.1320367	-0.78	0.434	-.3625144	.155779
z_polint_scale	0	(omitted)				
c.age#c.z_polint_scale	.0768305	.1664577	0.46	0.645	-.2498709	.4035319
1.male	-.261775	.0578573	-4.52	0.000	-.3753306	-.1482195
z_polint_scale	0	(omitted)				
male#c.z_polint_scale						
1	-.0577679	.0720259	-0.80	0.423	-.1991304	.0835947
race						
1	-.2140971	.1281296	-1.67	0.095	-.4655841	.03739
2	-.0942641	.1587976	-0.59	0.553	-.4059405	.2174123
3	-.1540502	.1565879	-0.98	0.325	-.4613958	.1532955
z_polint_scale	0	(omitted)				

race#c.z_polint_scale						
1	-.1132207	.1419257	-0.80	0.425	-.3917788	.1653374
2	-.06528	.1866928	-0.35	0.727	-.4317089	.301149
3	-.0028015	.1683261	-0.02	0.987	-.3331863	.3275833
education	.2540726	.1327028	1.91	0.056	-.0063839	.514529
z_polint_scale	0	(omitted)				
c.education#c.z_polint_scale	.0129663	.1533913	0.08	0.933	-.2880906	.3140232
income	.3398573	.1112942	3.05	0.002	.121418	.5582967
z_polint_scale	0	(omitted)				
c.income#c.z_polint_scale	-.1279431	.1225428	-1.04	0.297	-.368463	.1125768
5.sample_num	-.098646	.0658687	-1.50	0.135	-.2279249	.0306329
_cons	.4003501	.1530284	2.62	0.009	.0999984	.7007017

45 . mi estimate: regress z_dv_behavior c.z_latentnsc2##c.z_polknow_scale c.age##c.z_polknow_scale i.male##c.z_polknow_scal
> ation##c.z_polknow_scale c.income##c.z_polknow_scale i.sample_num

Multiple-imputation estimates	Imputations	=	25
Linear regression	Number of obs	=	906
	Average RVI	=	0.0105
	Largest FMI	=	0.0227
	Complete DF	=	887
DF adjustment: Small sample	DF: min	=	849.41
	avg	=	871.53
	max	=	884.29
Model F test: Equal FMI	F(18, 884.7)	=	3.96
Within VCE type: OLS	Prob > F	=	0.0000

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.0940267	.0323277	-2.91	0.004	-.1574747	-.0305788
z_polknow_scale	-.0703091	.15404	-0.46	0.648	-.37264	.2320218
c.z_latentnsc2#c.z_polknow_scale	-.0043937	.0371916	-0.12	0.906	-.0773879	.0686005
age	.0374415	.1433617	0.26	0.794	-.2439292	.3188122
z_polknow_scale	0	(omitted)				
c.age#c.z_polknow_scale	-.3282959	.1541988	-2.13	0.034	-.6309368	-.0256551
1.male	-.2629065	.0613562	-4.28	0.000	-.383328	-.142485
z_polknow_scale	0	(omitted)				
male#c.z_polknow_scale						
1	.0905027	.0674717	1.34	0.180	-.0419219	.2229273
race						
1	-.2275437	.1197366	-1.90	0.058	-.462551	.0074636
2	-.0256642	.1510773	-0.17	0.865	-.3221831	.2708546
3	-.1086219	.1475787	-0.74	0.462	-.398279	.1810351
z_polknow_scale	0	(omitted)				
race#c.z_polknow_scale						
1	.0986136	.1285233	0.77	0.443	-.1536365	.3508636
2	.1351741	.1623681	0.83	0.405	-.1835033	.4538515
3	.0639505	.1628908	0.39	0.695	-.2557617	.3836628
education	.1882438	.1352705	1.39	0.164	-.07725	.4537377
z_polknow_scale	0	(omitted)				

c.education#c.z_polknow_scale	**.1859171**	**.1487077**	**1.25**	**0.212**	**-.1059513**	**.4777856**
income	**.3143823**	**.1156349**	**2.72**	**0.007**	**.0874187**	**.5413458**
z_polknow_scale	**0**	(omitted)				
c.income#c.z_polknow_scale	**.0750527**	**.1161276**	**0.65**	**0.518**	**-.1528757**	**.302981**
5.sample_num	**-.0704387**	**.0675736**	**-1.04**	**0.298**	**-.2030633**	**.0621859**
_cons	**.3271119**	**.1467103**	**2.23**	**0.026**	**.0391659**	**.6150579**

```
46 .
47 .
end of do-file

48 .
```

[redacted]

```
24 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000008.tmp"

25 .
26 . ***** SM8 - Within-Sample Attitude-Behavior Comparisons *****
27 . mi estimate: regress z_dv_concern c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
   >  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num if (sample == "MTurk" | sample == "Bovitz

   Multiple-imputation estimates                   Imputations       =         25
   Linear regression                               Number of obs     =        893
                                                   Average RVI       =     0.0041
                                                   Largest FMI       =     0.0128
                                                   Complete DF       =        874
   DF adjustment:   Small sample                   DF:     min       =     855.82
                                                           avg       =     866.84
                                                           max       =     871.56
   Model F test:       Equal FMI                   F(  18,  872.0)   =       2.72
   Within VCE type:          OLS                   Prob > F          =     0.0001
```

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.0521632	.0533189	-0.98	0.328	-.1568117	.0524853
z_polengage_scale	.0241601	.2087403	0.12	0.908	-.3855357	.4338558
c.z_auth_scale#c.z_polengage_scale	-.1445166	.0700571	-2.06	0.039	-.2820171	-.0070162
age	.0232859	.1724169	0.14	0.893	-.3151152	.3616869
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.2050126	.2094076	-0.98	0.328	-.6160154	.2059901
1.male	-.156397	.0731192	-2.14	0.033	-.2999073	-.0128866
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.0996571	.0912445	1.09	0.275	-.0794277	.2787419
race						
1	-.1828713	.1517518	-1.21	0.229	-.4807141	.1149716
2	.2788096	.1906733	1.46	0.144	-.095426	.6530451
3	.0208637	.1851033	0.11	0.910	-.3424405	.3841678
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.102826	.1838586	0.56	0.576	-.2580355	.4636875
2	-.1414604	.2394287	-0.59	0.555	-.6113901	.3284694
3	-.0897341	.2235007	-0.40	0.688	-.5284008	.3489325
education	.3980507	.1631137	2.44	0.015	.0779054	.7181959
z_polengage_scale	0	(omitted)				

c.education#c.z_polengage_scale	.1059241	.1971776	0.54	0.591	-.2810817	.49293
income	-.2823272	.1376911	-2.05	0.041	-.552576	-.0120784
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0803977	.1527498	-0.53	0.599	-.3802059	.2194104
5.sample_num	-.0800779	.0844621	-0.95	0.343	-.2458514	.0856955
_cons	-.1738752	.1856793	-0.94	0.349	-.5383075	.1905571

28 . mi estimate: regress z_dv_behavior c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
> e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num if (sample == "MTurk" | sample == "Bovit

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 893
Average RVI = 0.0043
Largest FMI = 0.0221
Complete DF = 874
DF adjustment: Small sample — DF: min = 838.33
avg = 866.55
max = 871.80
Model F test: Equal FMI — F(18, 872.0) = 3.60
Within VCE type: OLS — Prob > F = 0.0000

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.0929786	.0426294	-2.18	0.029	-.1766469	-.0093103
z_polengage_scale	.1352129	.1667956	0.81	0.418	-.192157	.4625828
c.z_auth_scale#c.z_polengage_scale	-.0667652	.056012	-1.19	0.234	-.1766993	.0431689
age	-.1648374	.1378522	-1.20	0.232	-.4353985	.1057236
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1569554	.1674368	-0.94	0.349	-.4855822	.1716714
1.male	-.1914811	.058462	-3.28	0.001	-.3062239	-.0767383
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.0086631	.072954	0.12	0.906	-.1345231	.1518492
race						
1	-.2309742	.1213183	-1.90	0.057	-.469085	.0071367
2	-.0356739	.1522778	-0.23	0.815	-.334549	.2632012
3	-.1175847	.1478676	-0.80	0.427	-.4078049	.1726354
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0475959	.1468349	0.32	0.746	-.2405972	.3357891
2	.1232353	.191227	0.64	0.519	-.2520867	.4985573
3	.0863604	.1789786	0.48	0.630	-.2649245	.4376454
education	.2780955	.130378	2.13	0.033	.0222014	.5339895
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	-.0000936	.1574433	-0.00	1.000	-.3091097	.3089226
income	.3340454	.1102562	3.03	0.003	.1176421	.5504486
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0754688	.1227178	-0.61	0.539	-.3163391	.1654014
5.sample_num	-.0864887	.0675224	-1.28	0.201	-.2190145	.046037

_cons	.3529783	.148478	2.38	0.018	.061561	.6443955

```
29 .
30 . mi estimate: regress z_dv_concern c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengage
>  c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num if (sample == "MTurk" | sample == "Bovitz
```

Multiple-imputation estimates — Imputations = 25
Linear regression — Number of obs = 893
Average RVI = 0.0041
Largest FMI = 0.0132
Complete DF = 874
DF adjustment: **Small sample** — DF: min = 855.30
avg = 866.86
max = 871.71
Model F test: **Equal FMI** — F(18, 872.0) = 2.77
Within VCE type: **OLS** — Prob > F = 0.0001

z_dv_concern	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	-.0473112	.0382627	-1.24	0.217	-.1224091	.0277866
z_polengage_scale	.0367927	.2086061	0.18	0.860	-.3726398	.4462252
c.z_latentnsc2#c.z_polengage_scale	-.0999538	.0497011	-2.01	0.045	-.1975018	-.0024059
age	.0406518	.1737174	0.23	0.815	-.3003016	.3816052
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.1932957	.2105946	-0.92	0.359	-.6066281	.2200366
1.male	-.1549776	.073041	-2.12	0.034	-.2983344	-.0116208
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	.0955258	.091165	1.05	0.295	-.0834029	.2744545
race						
1	-.1874453	.1517272	-1.24	0.217	-.4852397	.1103492
2	.2801262	.1905583	1.47	0.142	-.0938836	.6541359
3	.0197855	.1849812	0.11	0.915	-.3432791	.38285
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0994073	.1838437	0.54	0.589	-.2614249	.4602394
2	-.1370074	.2397475	-0.57	0.568	-.6075631	.3335482
3	-.088268	.2234508	-0.40	0.693	-.5268367	.3503008
education	.3889072	.1629394	2.39	0.017	.0691043	.70871
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.1073882	.1968219	0.55	0.585	-.2789194	.4936958
income	-.2814471	.1376345	-2.04	0.041	-.551585	-.0113093
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0751319	.1526091	-0.49	0.623	-.374664	.2244003
5.sample_num	-.0794096	.0844101	-0.94	0.347	-.2450809	.0862617
_cons	-.1657927	.1852686	-0.89	0.371	-.5294189	.1978335

```
31 . mi estimate: regress z_dv_behavior c.z_latentnsc2##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
> e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num if (sample == "MTurk" | sample == "Bovit
```

Multiple-imputation estimates	Imputations	=	**25**
Linear regression	Number of obs	=	**893**
	Average RVI	=	**0.0042**
	Largest FMI	=	**0.0220**
	Complete DF	=	**874**
DF adjustment: **Small sample**	DF: min	=	**838.64**
	avg	=	**866.64**
	max	=	**871.81**
Model F test: **Equal FMI**	F(**18**, **872.0**)	=	**3.71**
Within VCE type: **OLS**	Prob > F	=	**0.0000**

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_latentnsc2	**-.0829208**	**.0305763**	**-2.71**	**0.007**	**-.1429325**	**-.0229091**
z_polengage_scale	**.1478958**	**.1665811**	**0.89**	**0.375**	**-.1790531**	**.4748447**
c.z_latentnsc2#c.z_polengage_scale	**-.031185**	**.0397161**	**-0.79**	**0.433**	**-.1091355**	**.0467654**
age	**-.135783**	**.1388145**	**-0.98**	**0.328**	**-.4082329**	**.1366669**
z_polengage_scale	**0**	(omitted)				
c.age#c.z_polengage_scale	**-.1646872**	**.1682885**	**-0.98**	**0.328**	**-.4949858**	**.1656113**
1.male	**-.1892882**	**.058365**	**-3.24**	**0.001**	**-.3038406**	**-.0747357**
z_polengage_scale	**0**	(omitted)				
male#c.z_polengage_scale						
1	**.0059136**	**.0728472**	**0.08**	**0.935**	**-.1370629**	**.1488901**
race						
1	**-.2337608**	**.1212276**	**-1.93**	**0.054**	**-.4716937**	**.0041721**
2	**-.0317246**	**.1520967**	**-0.21**	**0.835**	**-.3302442**	**.266795**
3	**-.1200929**	**.1476818**	**-0.81**	**0.416**	**-.4099483**	**.1697625**
z_polengage_scale	**0**	(omitted)				
race#c.z_polengage_scale						
1	**.0411596**	**.1467407**	**0.28**	**0.779**	**-.2468485**	**.3291677**
2	**.1153455**	**.1913446**	**0.60**	**0.547**	**-.2602072**	**.4908981**
3	**.0832639**	**.1788169**	**0.47**	**0.642**	**-.2677035**	**.4342313**
education	**.2725266**	**.1301772**	**2.09**	**0.037**	**.0170267**	**.5280265**
z_polengage_scale	**0**	(omitted)				
c.education#c.z_polengage_scale	**-.0028494**	**.1570841**	**-0.02**	**0.986**	**-.3111607**	**.3054619**
income	**.3358328**	**.1101276**	**3.05**	**0.002**	**.119682**	**.5519837**
z_polengage_scale	**0**	(omitted)				
c.income#c.z_polengage_scale	**-.0711166**	**.1225049**	**-0.58**	**0.562**	**-.3115688**	**.1693356**
5.sample_num	**-.0864479**	**.0674403**	**-1.28**	**0.200**	**-.2188125**	**.0459167**
_cons	**.3608776**	**.148066**	**2.44**	**0.015**	**.070269**	**.6514863**

```
32 .
33 . mi estimate: regress z_dv_restrictions c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_pole
   > scale c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num if (sample == "Lucid2020" | sample =
   > .

Multiple-imputation estimates                 Imputations       =         25
Linear regression                             Number of obs     =      3,059
                                              Average RVI       =     0.0066
                                              Largest FMI       =     0.0327
                                              Complete DF       =       3040
DF adjustment:   Small sample                 DF:     min       =   2,600.75
                                                      avg       =   2,963.01
                                                      max       =   3,037.53
Model F test:       Equal FMI                 F(  18, 3036.9)   =      18.97
Within VCE type:          OLS                 Prob > F          =     0.0000
```

z_dv_restrictions	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.073524	.0176853	-4.16	0.000	-.1082003	-.0388476
z_polengage_scale	.2619254	.0708646	3.70	0.000	.1229776	.4008732
c.z_auth_scale#c.z_polengage_scale	-.0600664	.01699	-3.54	0.000	-.0933795	-.0267533
age	-.502108	.0823493	-6.10	0.000	-.6635791	-.3406368
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.3794657	.077676	-4.89	0.000	-.5317788	-.2271527
1.male	.0365026	.0359439	1.02	0.310	-.0339752	.1069804
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.076238	.0339233	-2.25	0.025	-.142754	-.0097221
race						
1	-.1047645	.0661977	-1.58	0.114	-.2345614	.0250324
2	.2490132	.0856557	2.91	0.004	.0810642	.4169622
3	.1082596	.0809436	1.34	0.181	-.0504503	.2669694
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	-.0154638	.0632375	-0.24	0.807	-.1394566	.108529
2	-.0024816	.0777574	-0.03	0.975	-.1549441	.1499808
3	-.0583891	.0742465	-0.79	0.432	-.2039675	.0871893
education	.2791486	.069316	4.03	0.000	.1432373	.4150598
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	.0113224	.0657689	0.17	0.863	-.1176345	.1402794
income	.2874172	.0633662	4.54	0.000	.1631716	.4116628
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0075237	.0622721	-0.12	0.904	-.1296303	.1145829
4.sample_num	-.2199811	.034871	-6.31	0.000	-.2883543	-.151608
_cons	.0285886	.0852937	0.34	0.738	-.1386512	.1958283

```
34 . mi estimate: regress z_dv_behavior c.z_auth_scale##c.z_polengage_scale c.age##c.z_polengage_scale i.male##c.z_polengag
> e c.education##c.z_polengage_scale c.income##c.z_polengage_scale i.sample_num if (sample == "Lucid2020" | sample == "L
> .

Multiple-imputation estimates                Imputations       =         25
Linear regression                            Number of obs     =      3,059
                                             Average RVI       =     0.0132
                                             Largest FMI       =     0.0588
                                             Complete DF       =       3040
DF adjustment:   Small sample                DF:     min       =   2,032.83
                                                     avg       =   2,862.56
                                                     max       =   3,037.06
Model F test:       Equal FMI                F(  18, 3033.8)   =      37.40
Within VCE type:          OLS                Prob > F          =     0.0000
```

z_dv_behavior	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
z_auth_scale	-.0519856	.0189907	-2.74	0.006	-.0892215	-.0147496
z_polengage_scale	.3502302	.0761514	4.60	0.000	.2009161	.4995443
c.z_auth_scale#c.z_polengage_scale	-.0361594	.0182571	-1.98	0.048	-.071957	-.0003618
age	.1028497	.0884554	1.16	0.245	-.0705947	.2762941
z_polengage_scale	0	(omitted)				
c.age#c.z_polengage_scale	-.4868012	.0845312	-5.76	0.000	-.6525781	-.3210243
1.male	-.0561556	.038667	-1.45	0.147	-.1319733	.0196621
z_polengage_scale	0	(omitted)				
male#c.z_polengage_scale						
1	-.0406449	.036917	-1.10	0.271	-.113037	.0317472
race						
1	-.1352505	.0710772	-1.90	0.057	-.2746148	.0041138
2	.2280047	.0919453	2.48	0.013	.0477233	.408286
3	.0657699	.0869084	0.76	0.449	-.1046355	.2361752
z_polengage_scale	0	(omitted)				
race#c.z_polengage_scale						
1	.0471747	.067913	0.69	0.487	-.0859855	.180335
2	-.02896	.0834946	-0.35	0.729	-.1926716	.1347517
3	-.1254546	.0797838	-1.57	0.116	-.2818904	.0309812
education	.2429185	.0744113	3.26	0.001	.0970168	.3888202
z_polengage_scale	0	(omitted)				
c.education#c.z_polengage_scale	-.0414871	.0709938	-0.58	0.559	-.1806914	.0977172
income	.2767759	.0681223	4.06	0.000	.1432042	.4103476
z_polengage_scale	0	(omitted)				
c.income#c.z_polengage_scale	-.0548713	.0677963	-0.81	0.418	-.187827	.0780844
4.sample_num	-.6605303	.0374406	-17.64	0.000	-.7339418	-.5871188
_cons	.151294	.0915946	1.65	0.099	-.0283003	.3308882

```
35 .
36 .
end of do-file

37 .
```

```
34 .
35 .
   end of do-file

36 . do "C:\Users\14258\AppData\Local\Temp\STD24bc0_000009.tmp"

37 .
38 .
39 . ***** SM3 - Controls for PID/IDEO for Mediation Test *****
40 . import excel "pooled_data.xlsx", sheet("Sheet1") firstrow clear // run this separately since the MI is different
   (190 vars, 10,223 obs)

41 . mi set mlong

42 . mi register imputed age education income male race sample_num pid ideo // add pid and ideo so no extra listwise deleti
   > being compared use fixed observations)
   (1556 m=0 obs now marked as incomplete)

43 . mi impute chained (logit) male (mlogit) race sample_num (regress) age education income pid ideo, add(25)
   note: variable sample_num contains no soft missing (.) values; imputing nothing

   Conditional models:
            education: regress education i.sample_num i.race i.male age ideo pid income
                 race: mlogit race i.sample_num education i.male age ideo pid income
                 male: logit male i.sample_num education i.race age ideo pid income
                  age: regress age i.sample_num education i.race i.male ideo pid income
                ideo: regress ideo i.sample_num education i.race i.male age pid income
                  pid: regress pid i.sample_num education i.race i.male age ideo income
               income: regress income i.sample_num education i.race i.male age ideo pid

   Performing chained iterations ...

   Multivariate imputation                     Imputations =       25
   Chained equations                                 added =       25
   Imputed: m=1 through m=25                       updated =        0

   Initialization: monotone                     Iterations =      250
                                                   burn-in =       10

                 male: logistic regression
                 race: multinomial logistic regression
           sample_num: multinomial logistic regression
                  age: linear regression
            education: linear regression
               income: linear regression
                  pid: linear regression
                ideo: linear regression
```

	Observations per *m*			
Variable	Complete	Incomplete	Imputed	Total
male	**9681**	**542**	**542**	**10223**
race	**9898**	**325**	**325**	**10223**
sample_num	**10223**	**0**	**0**	**10223**
age	**9666**	**557**	**557**	**10223**
education	**9926**	**297**	**297**	**10223**
income	**9075**	**1148**	**1148**	**10223**
pid	**9462**	**761**	**761**	**10223**
ideo	**9520**	**703**	**703**	**10223**

(Complete + Incomplete = Total; Imputed is the minimum across *m*
 of the number of filled-in observations.)

```
44 .
45 . *Concern, Authoritarianism
46 . quietly mi estimate: regress dv_concern c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_s
   > ducation##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51 pid ideo

47 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

48 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.152566	.0451916	3.38	0.001	.0639723	.2411598

```
49 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0924512	.0183537	-5.04	0.000	-.1284318	-.0564707

```
50 . *Restrictions, Authoritarianism
51 . quietly mi estimate: regress dv_restrictions c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polen
   > e c.education##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51 pid ideo

52 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

53 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	.1806922	.0292943	6.17	0.000	.1232662	.2381183

```
54 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	-.0890221	.0139113	-6.40	0.000	-.1162926	-.0617517

```
55 . *Behavior, Authoritarianism
56 . quietly mi estimate: regress dv_behavior c.auth_scale##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_
   > education##c.polengage_scale c.income##c.polengage_scale i.sample_num state_1-state_51 pid ideo
```

```
57 . quietly mimrgns, at(auth_scale=(0 1) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

58 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.0260351**	**.0351549**	**0.74**	**0.459**	**-.0428884**	**.0949585**

```
59 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.0157371**	**.0213824**	**0.74**	**0.462**	**-.0261846**	**.0576587**

```
60 . *Concern, Latent NSC
61 . quietly mi estimate: regress dv_concern c.latentnsc2##c.polengage_scale c.age##c.polengage_scale i.male##c.polengage_s
   > ducation##c.polengage_scale c.income##c.polengage_scale i.sample_num pid ideo

62 . quietly mimrgns, at(latentnsc2=( 0.055  0.782) polengage_scale=(0 .25 .5 .75 1)) coeflegend post

63 .     lincom _b[6._at]-_b[1._at]

 ( 1)  - 1bn._at + 6._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.1602181**	**.1658997**	**0.97**	**0.334**	**-.1653569**	**.4857931**

```
64 .     lincom _b[10._at]-_b[5._at]

 ( 1)  - 5._at + 10._at = 0
```

	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
(1)	**.049381**	**.0688066**	**0.72**	**0.473**	**-.0856506**	**.1844125**

```
65 .
66 . clear // run this separately since the MI is different

67 .
68 .
   end of do-file

69 .
```