

name: Mirella Charros

log: /Users/mirellacharos/Documents/ECO722\_Project/Data/Project\_1\_Charros.smcl

log type: smcl

opened on: 25 May 2025, 11:14:50

. \* Defining treatment group and control groups

. generate expansion\_state = 0

. replace expansion\_state = 1 if inlist(\_state, 4, 5, 6, 8, 9, 10, 11, 15, 17, 19, 21, 24, 25, 27, 32, 34, 35, 36, 38, 39, 41, 44, 50, 53, 54)

(1,979,261 real changes made)

. generate post2014 = iyear >= 2014 // post 2014

. label variable post2014 "Post-period (1 = 2014 and after)"

. generate DiD = expansion\_state \* post2014 // interaction term

. label variable DiD "Interaction: Expansion × Post-2014"

. tab \_state expansion\_state if iyear == 2014 // check

| expansion\_state

STATE FIPS CODE | 0 1. | Total

-----+-----+-----

Alabama	8,625	0		8,625
Alaska	4,388	0		4,388
Arizona	0	14,855		14,855
Arkansas	0	5,253		5,253
California	0	8,226		8,226
Colorado	0	12,862		12,862
Connecticut	0	7,950		7,950
Delaware	0	4,300		4,300
District of Columbia	0	4,074		4,074

Florida	9,809	0	9,809
Georgia	6,348	0	6,348
Hawaii	0	7,086	7,086
Idaho	5,480	0	5,480
Illinois	0	4,911	4,911
Indiana	11,516	0	11,516
Iowa	0	8,121	8,121
Kansas	13,736	0	13,736
Kentucky	0	11,184	11,184
Louisiana	6,771	0	6,771
Maine	9,137	0	9,137
Maryland	0	12,567	12,567
Massachusetts	0	15,652	15,652
Michigan	8,461	0	8,461
Minnesota	0	16,416	16,416
Mississippi	3,966	0	3,966
Missouri	7,076	0	7,076
Montana	7,502	0	7,502
Nebraska	21,414	0	21,414
Nevada	0	3,697	3,697
New Hampshire	6,191	0	6,191
New Jersey	0	13,041	13,041
New Mexico	0	8,734	8,734
New York	0	6,837	6,837
North Carolina	7,134	0	7,134
North Dakota	0	7,782	7,782

Ohio	0	10,929	10,929
Oklahoma	7,740	0	7,740
Oregon	0	5,220	5,220
Pennsylvania	10,990	0	10,990
Rhode Island	0	6,450	6,450
South Carolina	11,025	0	11,025
South Dakota	7,400	0	7,400
Tennessee	4,593	0	4,593
Texas	15,403	0	15,403
Utah	15,003	0	15,003
Vermont	0	6,474	6,474
Virginia	9,463	0	9,463
Washington	0	10,082	10,082
West Virginia	0	6,199	6,199
Wisconsin	6,951	0	6,951
Wyoming	6,410	0	6,410
Guam	2,475	0	2,475
Puerto Rico	5,935	0	5,935
-----+-----+-----			
Total	240,942	218,902	459,844

. tab DiD

Interaction |

: Expansion |

× |

	Post-2014	Freq.	Percent	Cum.
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-----+-----

0	3,026,198	73.30	73.30
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1	1,102,152	26.70	100.00
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-----+-----

Total	4,128,350	100.00	
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. \* age restriction to 64 because of medicare coverage for elders

. drop if \_ageg5yr >= 10 // Since age group 10 = 65–69

(1,444,276 observations deleted)

. label define ageg5yr\_lbl ///

> 1 "18–24" ///

> 2 "25–29" ///

> 3 "30–34" ///

> 4 "35–39" ///

> 5 "40–44" ///

> 6 "45–49" ///

> 7 "50–54" ///

> 8 "55–59" ///

> 9 "60–64"

. label values \_ageg5yr ageg5yr\_lbl

. label variable \_ageg5yr "5-Year Age Category"

. \* genhlth

. ologit genhlth i.expansion\_state##i.post2014 i.sex i.race2 i.income2 i.ed

> uca, vce(cluster \_state)

Iteration 0: Log pseudolikelihood = -3342835.8

Iteration 1: Log pseudolikelihood = -3153043.8

Iteration 2: Log pseudolikelihood = -3149397.9

Iteration 3: Log pseudolikelihood = -3149390.7

Iteration 4: Log pseudolikelihood = -3149390.7

Ordered logistic regression                      Number of obs = 2,321,942

Wald chi2(25) = 38731.83

Prob > chi2 = 0.0000

Log pseudolikelihood = -3149390.7                      Pseudo R2 = 0.0579

(Std. err. adjusted for 54 clusters in \_state)

-----						
		Robust				
genhlth		Coefficient	std. err.	z	P> z	[95% conf. interval]
-----+-----						
1.expansion_state		-.0431728	.035992	-1.20	0.230	-.1137159 .0273702
1.post2014		.0744012	.0120744	6.16	0.000	.0507359 .0980666
expansion_state#post2014						
1 1		.0267704	.0148212	1.81	0.071	-.0022786 .0558193
sex						
Female		-.0421376	.0081907	-5.14	0.000	-.058191 -.026084
race2						
Black NH		.151935	.0226299	6.71	0.000	.1075812 .1962888
Asian NH		.1535203	.0343431	4.47	0.000	.086209 .2208316

NH/PI NH	-.017655	.0556017	-0.32	0.751	-.1266323	.0913223
AI/AN NH	.2376864	.0332851	7.14	0.000	.1724488	.3029239
Other NH	.0852018	.0304262	2.80	0.005	.0255676	.144836
Multiracial NH	.2381414	.0164828	14.45	0.000	.2058357	.2704472
Hispanic	.0065427	.0339091	0.19	0.847	-.059918	.0730034
income2						
\$10k–14,999	.0303719	.0137595	2.21	0.027	.0034038	.0573401
\$15k–19,999	-.4128224	.0220762	-18.70	0.000	-.4560909	-.369554
\$20k–24,999	-.6975743	.0289716	-24.08	0.000	-.7543576	-.6407909
\$25k–34,999	-.9816382	.0321066	-30.57	0.000	-1.044566	-.9187104
\$35k–49,999	-1.216325	.0334061	-36.41	0.000	-1.2818	-1.15085
\$50k–74,999	-1.416611	.0364759	-38.84	0.000	-1.488102	-1.345119
\$75k or more	-1.753422	.0383566	-45.71	0.000	-1.828599	-1.678244
educa						
Grades 1–8	.3391743	.0634716	5.34	0.000	.2147723	.4635764
Some high school	.0122342	.061325	0.20	0.842	-.1079606	.132429
High school graduate ..	-.4285153	.0607793	-7.05	0.000	-.5476407	-.30939
Some college or tech ..	-.6109472	.0606358	-10.08	0.000	-.7297912	-.4921033
College graduate	-.9937612	.0597795	-16.62	0.000	-1.110927	-.8765956
-----+-----						
/cut1	-3.416808	.0602605			-3.534917	-3.2987
/cut2	-1.694221	.0572731			-1.806474	-1.581967
/cut3	-.0088359	.0569635			-.1204823	.1028106
/cut4	1.516011	.0608551			1.396737	1.635285

```

-----

.     eststo genhlth

.     margins expansion_state#post2014, predict(outcome(1))

Predictive margins                Number of obs = 2,321,942

Model VCE: Robust

Expression: Pr(genhlth==1), predict(outcome(1))

```

```

-----

      |      Delta-method
      |  Margin  std. err.   z   P>|z|   [95% conf. interval]
-----+-----
expansion_state#post2014 |
      0 0 | .1982662 .0030049  65.98  0.000   .1923767 .2041558
      0 1 | .1874189 .0032049  58.48  0.000   .1811374 .1937005
      1 0 | .2047571 .0044179  46.35  0.000   .1960983 .213416
      1 1 | .1897734 .0036877  51.46  0.000   .1825457 .1970011
-----

```

```

.     marginsplot, title("Probability of Excellent Health by Group") ylabel(, g
> rid)

Variables that uniquely identify margins: expansion_state post2014

. * physhlth

.     nbreg physhlth i.expansion_state##i.post2014 i.sex i.race2 i.income2 i.educa,
vce(cluster_state)

```

Fitting Poisson model:

Iteration 0: Log pseudolikelihood = -13319585

Iteration 1: Log pseudolikelihood = -13316113

Iteration 2: Log pseudolikelihood = -13316111

Iteration 3: Log pseudolikelihood = -13316111

Fitting constant-only model:

Iteration 0: Log pseudolikelihood = -5703606.4

Iteration 1: Log pseudolikelihood = -4271928

Iteration 2: Log pseudolikelihood = -4270574.1

Iteration 3: Log pseudolikelihood = -4270573.7

Fitting full model:

Iteration 0: Log pseudolikelihood = -4222862.8

Iteration 1: Log pseudolikelihood = -4216022.2

Iteration 2: Log pseudolikelihood = -4215711.9

Iteration 3: Log pseudolikelihood = -4215711.6

Negative binomial regression                      Number of obs = 2,299,420

Wald chi2(25) = 19135.24

Dispersion: mean                      Prob > chi2 = 0.0000

Log pseudolikelihood = -4215711.6                      Pseudo R2 = 0.0128

(Std. err. adjusted for 54 clusters in \_state)

-----

		Robust					
physhlth		Coefficient	std. err.	z	P> z	[95% conf. interval]	
-----+-----							
1.expansion_state		.0430097	.0301393	1.43	0.154	-.0160622	.1020816



1.post2014	.0558292	.009876	5.65	0.000	.0364726	.0751859
expansion_state#post2014						
1 1	.0097906	.0123185	0.79	0.427	-.0143533	.0339344
sex						
Female	.1585321	.0067911	23.34	0.000	.1452218	.1718424
race2						
Black NH	-.1527891	.0185441	-8.24	0.000	-.1891348	-.1164433
Asian NH	-.0569599	.030109	-1.89	0.059	-.1159725	.0020527
NH/PI NH	-.4266399	.0481152	-8.87	0.000	-.5209439	-.3323359
AI/AN NH	.0124962	.0434321	0.29	0.774	-.0726292	.0976217
Other NH	.1345457	.0329305	4.09	0.000	.0700031	.1990883
Multiracial NH	.2027441	.0499517	4.06	0.000	.1048406	.3006476
Hispanic	-.2427209	.0193359	-12.55	0.000	-.2806185	-.2048232
income2						
\$10k–14,999	-.0085531	.0085347	-1.00	0.316	-.0252809	.0081747
\$15k–19,999	-.2940332	.0143118	-20.54	0.000	-.3220837	-.2659826
\$20k–24,999	-.507333	.0199061	-25.49	0.000	-.5463483	-.4683178
\$25k–34,999	-.7611804	.0233932	-32.54	0.000	-.8070303	-.7153306
\$35k–49,999	-.9744286	.0240399	-40.53	0.000	-1.021546	-.9273113
\$50k–74,999	-1.170698	.0266302	-43.96	0.000	-1.222892	-1.118504
\$75k or more	-1.447293	.0267083	-54.19	0.000	-1.49964	-1.394945

educa						
Grades 1–8	.1735456	.0439692	3.95	0.000	.0873675	.2597237
Some high school	.1693751	.0349715	4.84	0.000	.1008322	.237918
High school graduate ..	-.0259602	.0337273	-0.77	0.441	-.0920645	.0401441
Some college or tech ..	-.0147723	.0326721	-0.45	0.651	-.0788085	.0492639
College graduate	-.2857863	.0325584	-8.78	0.000	-.3495995	-.2219731
_cons	2.165437	.0384006	56.39	0.000	2.090173	2.240701
-----+-----						
/lnalpha	1.886013	.0137297			1.859103	1.912923
-----+-----						
alpha	6.593031	.0905205			6.41798	6.772857
-----						

. eststo physhlth

. margins expansion\_state#post2014

Predictive margins                      Number of obs = 2,299,420

Model VCE: Robust

Expression: Predicted number of events, predict()

-----						
	Delta-method					
	Margin	std. err.	z	P> z	[95% conf. interval]	
-----+-----						
expansion_state#post2014						
0 0	3.707557	.0845032	43.87	0.000	3.541933	3.87318

0 1	3.920434	.0775023	50.58	0.000	3.768532	4.072336
1 0	3.870497	.0751306	51.52	0.000	3.723243	4.01775
1 1	4.132996	.082177	50.29	0.000	3.971932	4.29406

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. \* menthlth

. nbreg menthlth i.expansion\_state##i.post2014 i.sex i.race2 i.income2 i.educa, ///  
vce(cluster \_state)

Fitting Poisson model:

Iteration 0: Log pseudolikelihood = -13482019

Iteration 1: Log pseudolikelihood = -13480784

Iteration 2: Log pseudolikelihood = -13480783

Fitting constant-only model:

Iteration 0: Log pseudolikelihood = -5780531.9

Iteration 1: Log pseudolikelihood = -4411294.5

Iteration 2: Log pseudolikelihood = -4410418.7

Iteration 3: Log pseudolikelihood = -4410418.5

Fitting full model:

Iteration 0: Log pseudolikelihood = -4369662.5

Iteration 1: Log pseudolikelihood = -4364773.6

Iteration 2: Log pseudolikelihood = -4364635.1

Iteration 3: Log pseudolikelihood = -4364635

Negative binomial regression                      Number of obs = 2,301,342

Wald chi2(25) = 11695.76

Dispersion: mean                                      Prob > chi2 = 0.0000

Log pseudolikelihood = -4364635                      Pseudo R2 = 0.0104

(Std. err. adjusted for 54 cluster in \_state)

Robust						
menthlth	Coefficient	std. err.	z	P> z	[95% conf. interval]	
-----+-----						
1.expansion_state	.0773481	.0305132	2.53	0.011	.0175433	.1371529
1.post2014	.0941037	.0094674	9.94	0.000	.075548	.1126594
expansion_state#post2014						
1 1	-.0239098	.0133934	-1.79	0.074	-.0501604	.0023408
sex						
Female	.3315315	.0080429	41.22	0.000	.3157677	.3472954
7	.620622	.1636999	3.79	0.000	.299776	.941468
9	.8176922	.0963122	8.49	0.000	.6289237	1.006461
race2						
Black NH	-.1400733	.0209858	-6.67	0.000	-.1812047	-.098942
Asian NH	-.1313111	.0356216	-3.69	0.000	-.2011282	-.061494
NH/PI NH	-.3492496	.0302457	-11.55	0.000	-.40853	-.2899692
AI/AN NH	.0325143	.0415745	0.78	0.434	-.0489702	.1139988
Other NH	.1755937	.0253561	6.93	0.000	.1258967	.2252908
Multiracial NH	.2533668	.0499153	5.08	0.000	.1555346	.3511991
Hispanic	-.2045849	.0217639	-9.40	0.000	-.2472414	-.1619284
income2						

\$10k–14,999	-.0786235	.0081484	-9.65	0.000	-.094594	-.062653
\$15k–19,999	-.2923644	.0111876	-26.13	0.000	-.3142917	-.2704371
\$20k–24,999	-.4542589	.0142957	-31.78	0.000	-.482278	-.4262399
\$25k–34,999	-.67786	.0180861	-37.48	0.000	-.7133081	-.6424119
\$35k–49,999	-.8627713	.0206767	-41.73	0.000	-.9032969	-.8222456
\$50k–74,999	-1.041584	.0253684	-41.06	0.000	-1.091306	-.9918633
\$75k or more	-1.31421	.0253891	-51.76	0.000	-1.363972	-1.264448
educa						
Grades 1–8	.1975125	.0447661	4.41	0.000	.1097725	.2852524
Some high school	.3805839	.0488702	7.79	0.000	.2848001	.4763678
High school graduate ..	.1842981	.0490007	3.76	0.000	.0882585	.2803377
Some college or tech ..	.2320091	.0504529	4.60	0.000	.1331234	.3308949
College graduate	.0173351	.051197	0.34	0.735	-.0830091	.1176793
_cons	1.757334	.0581292	30.23	0.000	1.643402	1.871265
-----+-----						
/lnalpha	1.859504	.0169191			1.826343	1.892665
-----+-----						
alpha	6.420553	.10863			6.211134	6.637033
-----						

. eststo menthlth

. margins expansion\_state#post2014

Predictive margins                      Number of obs = 2,301,342

Model VCE: Robust

Expression: Predicted number of events, predict()

```
-----  
      |      Delta-method  
      |  Margin  std. err.   z   P>|z|   [95% conf. interval]  
-----+-----  
expansion_state#post2014 |  
      0 0 | 3.741213 .0817747  45.75  0.000   3.580938   3.901489  
      0 1 | 4.110372 .0932325  44.09  0.000   3.92764   4.293105  
      1 0 | 4.042074 .0807474  50.06  0.000   3.883813   4.200336  
      1 1 | 4.335999 .079411  54.60  0.000   4.180356   4.491641  
-----
```

```
. * effect on personal doctor access after expansion  
. generate has_doc = . // create binary variable  
(2,684,074 missing values generated)  
. replace has_doc = 1 if inlist(persdoc2, 1, 2)  
(2,139,236 real changes made)  
. replace has_doc = 0 if persdoc2 == 3  
(534,815 real changes made)  
. label var has_doc "Has a personal doctor"  
. logit has_doc i.expansion_state##i.post2014 /// i.sex i.race2 i.income2 i.educa,  
vce(cluster _state)  
  
Iteration 0: Log pseudolikelihood = -1143280  
Iteration 1: Log pseudolikelihood = -1078651.7  
Iteration 2: Log pseudolikelihood = -1075924.7
```

Iteration 3: Log pseudolikelihood = -1075921.1

Iteration 4: Log pseudolikelihood = -1075921.1

Logistic regression                      Number of obs = 2,320,108

Wald chi2(25) = 13173.61

Prob > chi2 = 0.0000

Log pseudolikelihood = -1075921.1                      Pseudo R2 = 0.0589

(Std. err. adjusted for 54 clusters in \_state)

-----						
	Robust					
has_doc	Coefficient	std. err.	z	P> z	[95% conf. interval]	
-----+-----						
1.expansion_state	.0753967	.0975831	0.77	0.440	-.1158626	.2666561
1.post2014	-.1494341	.0218905	-6.83	0.000	-.1923386	-.1065296
expansion_state#post2014						
1 1	.0612446	.0470462	1.30	0.193	-.0309643	.1534535
sex						
Female	.7519308	.0141646	53.09	0.000	.7241688	.7796928
race2						
Black NH	.0469354	.045777	1.03	0.305	-.0427859	.1366568
Asian NH	-.4704467	.0958128	-4.91	0.000	-.6582363	-.282657
NH/PI NH	-.4275635	.1395258	-3.06	0.002	-.701029	-.154098
AI/AN NH	-.4607692	.1004924	-4.59	0.000	-.6577306	-.2638078

```

Other NH | -.3395206 .0445281 -7.62 0.000 -.426794 -.2522471
Multiracial NH | -.2129547 .065192 -3.27 0.001 -.3407286 -.0851808
Hispanic | -.5789076 .108045 -5.36 0.000 -.7906719 -.3671432
|
income2 |
$10k–14,999 | .0903289 .0328353 2.75 0.006 .0259729 .1546848
$15k–19,999 | -.0830251 .0440277 -1.89 0.059 -.1693178 .0032676
$20k–24,999 | -.0287611 .0528184 -0.54 0.586 -.1322834 .0747611
$25k–34,999 | .1036107 .0589657 1.76 0.079 -.01196 .2191813
$35k–49,999 | .3318571 .0640443 5.18 0.000 .2063327 .4573815
$50k–74,999 | .5799757 .0682417 8.50 0.000 .4462244 .713727
$75k or more | .8737503 .071637 12.20 0.000 .7333444 1.014156
|
educa |
Grades 1–8 | .216975 .0534479 4.06 0.000 .112219 .3217311
Some high school | .2633402 .0595948 4.42 0.000 .1465366 .3801438
High school graduate .. | .5030027 .0636899 7.90 0.000 .3781729 .6278326
Some college or tech .. | .5736002 .0644433 8.90 0.000 .4472937 .6999066
College graduate | .6369738 .0688567 9.25 0.000 .5020171 .7719304
|
_cons | .2042412 .0770084 2.65 0.008 .0533075 .3551748

```

```
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```

```

. eststo has_doc

```

```

. margins expansion_state#post2014

```

```

Predictive margins

```

```

Number of obs = 2,320,108

```



Expression:  $\Pr(\text{has\_doc})$ , `predict()`

		Delta-method					
		Margin	std. err.	z	P> z	[95% conf. interval]	
-----+-----							
expansion_state#post2014							
0 0		.8095014	.0087406	92.61	0.000	.7923702 .8266326	
0 1		.7869191	.009549	82.41	0.000	.7682033 .8056348	
1 0		.8202132	.0106	77.38	0.000	.7994377 .8409888	
1 1		.8076386	.0106129	76.10	0.000	.7868377 .8284395	

```
. margins, dydx(post2014) at(expansion_state=(0 1))
```

Average marginal effects      Number of obs = 2,320,108

Model VCE: Robust

Expression:  $\Pr(\text{has\_doc})$ , `predict()`

dy/dx wrt: 1.post2014

1.\_at: expansion\_state = 0

2.\_at: expansion\_state = 1

	Delta-method				
	dy/dx	std. err.	z	P> z	[95% conf. interval]

0.post2014 | (base outcome)

\_at |

2 | -.0125747 .0058827 -2.14 0.033 -.0241045 -.0010448

Prob > chi2 = 0.0000

Log pseudolikelihood = -861761.18

Pseudo R2 = 0.1133

(Std. err. adjusted for 54 clusters in \_state)

-----						
Robust						
couldnt_afford   Coefficient std. err. z P> z  [95% conf. interval]						
-----+-----						
1.expansion_state	-.1097654	.0715752	-1.53	0.125	-.2500502	.0305194
1.post2014	-.104354	.0144806	-7.21	0.000	-.1327354	-.0759726
expansion_state#post2014						
1 1	-.1026177	.0371707	-2.76	0.006	-.1754709	-.0297645
sex						
Female	.2668806	.0169051	15.79	0.000	.2337472	.300014
7	.2049503	.2421112	0.85	0.397	-.2695789	.6794795
9	.6366025	.165036	3.86	0.000	.3131379	.9600671
race2						
Black NH	.0717721	.0361149	1.99	0.047	.0009881	.1425561
Asian NH	-.0606286	.0585742	-1.04	0.301	-.1754319	.0541747
NH/PI NH	-.0637055	.0864226	-0.74	0.461	-.2330906	.1056797
AI/AN NH	.0781763	.0626731	1.25	0.212	-.0446606	.2010133
Other NH	.3603603	.0438774	8.21	0.000	.2743621	.4463584
Multiracial NH	.2808812	.0834038	3.37	0.001	.1174128	.4443496
Hispanic	.1926317	.0978034	1.97	0.049	.0009405	.3843228

```

      |
income2 |
$10k–14,999 | .048865 .0174653 2.80 0.005 .0146337 .0830963
$15k–19,999 | .0084293 .023813 0.35 0.723 -.0382433 .055102
$20k–24,999 | -.1239651 .0310895 -3.99 0.000 -.1848994 -.0630308
$25k–34,999 | -.4615728 .0395587 -11.67 0.000 -.5391064 -.3840392
$35k–49,999 | -.8554864 .0466061 -18.36 0.000 -.9468326 -.7641402
$50k–74,999 | -1.359619 .0503118 -27.02 0.000 -1.458228 -1.26101
$75k or more | -2.1287 .0488028 -43.62 0.000 -2.224352 -2.033049
      |
educat |
Grades 1–8 | -.1146752 .0416324 -2.75 0.006 -.1962732 -.0330772
Some high school | -.0919201 .0585423 -1.57 0.116 -.2066608 .0228207
High school graduate .. | -.3592432 .0657867 -5.46 0.000 -.4881828 -.2303036
Some college or tech .. | -.2743842 .069998 -3.92 0.000 -.4115777 -.1371907
College graduate | -.5092904 .0709971 -7.17 0.000 -.6484422 -.3701387
      |
      _cons | -.6004719 .0723276 -8.30 0.000 -.7422313 -.4587125

```

```
-----
```

```
.      eststo couldnt_afford
```

```
.      margins expansion_state#post2014
```

```
Predictive margins          Number of obs = 2,322,062
```

```
Model VCE: Robust
```

```
Expression: Pr(couldnt_afford), predict()
```



_at						
1	-.0124045	.0018782	-6.60	0.000	-.0160857	-.0087234
2	-.0224231	.0040388	-5.55	0.000	-.0303391	-.0145072

-----

Note: dy/dx for factor levels is the discrete change from the base level.

```
. * effect on checkup1
. drop if checkup1 > 4 // Optional if you want to restrict to clean ordinal levels
(60,531 observations deleted)

. ologit checkup1 i.expansion_state##i.post2014 // i.sex i.race2 i.income2 i.educa,
vce(cluster _state)

Iteration 0: Log pseudolikelihood = -2172776
Iteration 1: Log pseudolikelihood = -2139593.1
Iteration 2: Log pseudolikelihood = -2139258.9
Iteration 3: Log pseudolikelihood = -2139258.7
```

Ordered logistic regression                      Number of obs = 2,281,082

Wald chi2(25) = 13598.31

Prob > chi2 = 0.0000

Log pseudolikelihood = -2139258.7                      Pseudo R2 = 0.0154

(Std. err. adjusted for 54 clusters in \_state)

-----

	Robust				
checkup1	Coefficient	std. err.	z	P> z	[95% conf. interval]

```

-----+-----
1.expansion_state | -.0799761 .0842731 -0.95 0.343 -.2451483 .0851961
1.post2014 | -.1245885 .0271695 -4.59 0.000 -.1778398 -.0713371
|
expansion_state#post2014 |
1 1 | -.0113015 .050258 -0.22 0.822 -.1098053 .0872024
|
sex |
Female | -.4785534 .0132043 -36.24 0.000 -.5044333 -.4526735 |
race2 |
Black NH | -.7556857 .0339224 -22.28 0.000 -.8221723 -.689199
Asian NH | -.0967577 .0469721 -2.06 0.039 -.1888214 -.004694
NH/PI NH | -.0646199 .0295768 -2.18 0.029 -.1225894 -.0066504
AI/AN NH | -.0607339 .0413084 -1.47 0.141 -.1416968 .0202291
Other NH | -.1618051 .0289555 -5.59 0.000 -.2185567 -.1050534
Multiracial NH | .0977644 .0346951 2.82 0.005 .0297632 .1657656
Hispanic | -.0793276 .0852505 -0.93 0.352 -.2464156 .0877603
|
income2 |
$10k-14,999 | -.0245684 .0172701 -1.42 0.155 -.0584172 .0092804
$15k-19,999 | .094882 .0213651 4.44 0.000 .0530072 .1367568
$20k-24,999 | .0879831 .0276043 3.19 0.001 .0338798 .1420865
$25k-34,999 | .0225507 .0310337 0.73 0.467 -.0382743 .0833756
$35k-49,999 | -.1229222 .033945 -3.62 0.000 -.1894532 -.0563913
$50k-74,999 | -.2836085 .0369801 -7.67 0.000 -.3560882 -.2111288
$75k or more | -.4501706 .0393602 -11.44 0.000 -.5273152 -.3730261

```

educa							
Grades 1–8	-.0105	.0486274	-0.22	0.829	-.1058079	.0848078	
Some high school	.0460265	.049032	0.94	0.348	-.0500744	.1421274	
High school graduate ..	-.0601456	.0555752	-1.08	0.279	-.169071	.0487798	
Some college or tech ..	-.052279	.0578248	-0.90	0.366	-.1656135	.0610555	
College graduate	-.0754704	.0600421	-1.26	0.209	-.1931507	.0422099	

-----+-----						
/cut1	.0963819	.0687835		-.0384314	.2311951	
/cut2	.8973267	.0724635		.7553009	1.039353	
/cut3	1.724722	.0766487		1.574493	1.87495	

```

.      eststo checkup1
.      margins expansion_state#post2014, predict(outcome(1))
Predictive margins                Number of obs = 2,281,082
Model VCE: Robust
Expression: Pr(checkup1==1), predict(outcome(1))

```

-----+-----						
		Delta-method				
		Margin	std. err.	z	P> z	[95% conf. interval]
-----+-----						
expansion_state#post2014						
0 0	.6649524	.0107955	61.60	0.000	.6437935	.6861113
0 1	.6913976	.0086337	80.08	0.000	.6744759	.7083193



1 0	.6820564	.0140421	48.57	0.000	.6545344	.7095785
1 1	.7100378	.0101031	70.28	0.000	.690236	.7298396

-----

. \* majpr health outcomes- diabetes

. logit diabetes i.expansion\_state##i.post2014 /// i.sex i.race2 i.income2 i.educa,  
vce(cluster \_state)

Iteration 0: Log pseudolikelihood = -699354.02

Iteration 1: Log pseudolikelihood = -677637.04

Iteration 2: Log pseudolikelihood = -675278.68

Iteration 3: Log pseudolikelihood = -675273.42

Iteration 4: Log pseudolikelihood = -675273.42

Logistic regression                      Number of obs = 2,253,076

Wald chi2(25) = 17178.61

Prob > chi2 = 0.0000

Log pseudolikelihood = -675273.42                      Pseudo R2 = 0.0344

(Std. err. adjusted for 54 clusters in \_state)

-----

		Robust	
diabetes		Coefficient std. err.	z P> z  [95% conf. interval]

-----+-----

1.expansion_state		-.0720328	.0437371	-1.65	0.100	-.1577558	.0136903
-------------------	--	-----------	----------	-------	-------	-----------	----------

1.post2014		.0672246	.016706	4.02	0.000	.0344814	.0999679
------------	--	----------	---------	------	-------	----------	----------

|

expansion\_state#post2014 |

1 1 | .0385373 .0222314 1.73 0.083 -.0050356 .0821101

|

sex |

Female | -.1097343 .0107472 -10.21 0.000 -.1307985 -.0886701

|

race2 |

Black NH | .4198008 .0276121 15.20 0.000 .3656821 .4739195

Asian NH | .3084276 .0718598 4.29 0.000 .167585 .4492701

NH/PI NH | -.090091 .0905344 -1.00 0.320 -.2675352 .0873533

AI/AN NH | .3794379 .0660421 5.75 0.000 .2499977 .5088781

Other NH | .0877644 .0430471 2.04 0.041 .0033935 .1721352

Multiracial NH | .137103 .0278119 4.93 0.000 .0825927 .1916132

Hispanic | -.0909611 .0451581 -2.01 0.044 -.1794693 -.002453

|

income2 |

\$10k–14,999 | .1330148 .0150923 8.81 0.000 .1034344 .1625952

\$15k–19,999 | -.1297786 .0154876 -8.38 0.000 -.1601337 -.0994235

\$20k–24,999 | -.2956731 .0143329 -20.63 0.000 -.323765 -.2675812

\$25k–34,999 | -.4560238 .0162399 -28.08 0.000 -.4878533 -.4241943

\$35k–49,999 | -.550487 .0176392 -31.21 0.000 -.5850593 -.5159148

\$50k–74,999 | -.6752311 .0200255 -33.72 0.000 -.7144804 -.6359817

\$75k or more | -.9780043 .022927 -42.66 0.000 -1.02294 -.9330682

|

educa |

Grades 1–8 | -.0218049 .0594706 -0.37 0.714 -.1383651 .0947554

```

Some high school | -.3110377 .0582498 -5.34 0.000 -.4252051 -.1968702
High school graduate .. | -.4307314 .0581786 -7.40 0.000 -.5447594 -.3167034
Some college or tech .. | -.4645039 .0580742 -8.00 0.000 -.5783271 -.3506806
College graduate | -.7435657 .0594829 -12.50 0.000 -.86015 -.6269814
|
_cons | -1.212621 .0670217 -18.09 0.000 -1.343982 -1.081261

```

```

. eststo diabetes
. margins expansion_state#post2014

```

Predictive margins                      Number of obs = 2,253,076

Model VCE: Robust

Expression: Pr(diabetes), predict()

```

|      Delta-method
|  Margin  std. err.   z  P>|z|   [95% conf. interval]
-----+-----
expansion_state#post2014 |
0 0 | .0924909 .002243  41.23 0.000   .0880946 .0968871
0 1 | .0981508 .0024071  40.77 0.000   .0934329 .1028687
1 0 | .0867464 .0025651  33.82 0.000   .0817189 .091774
1 1 | .0952939 .0033559  28.40 0.000   .0887164 .1018714

```

```

. margins, dydx(post2014) at(expansion_state=(0 1))

```

Average marginal effects

Number of obs = 2,253,076

Model VCE: Robust

Expression: Pr(diabetes), predict()

dy/dx wrt: 1.post2014

1.\_at: expansion\_state = 0

2.\_at: expansion\_state = 1

-----						
		Delta-method				
		dy/dx	std. err.	z	P> z	[95% conf. interval]
-----+-----						
0.post2014		(base outcome)				
-----+-----						
1.post2014						
	_at					
1		.00566	.0014143	4.00	0.000	.0028879 .008432
2		.0085475	.0013117	6.52	0.000	.0059766 .0111184
-----						

Note: dy/dx for factor levels is the discrete change from the base level.

. \* heart attack

. logit heart\_attack i.expansion\_state##i.post2014 /// i.sex i.race2 i.income2 i.educa,  
vce(cluster \_state)

Iteration 5: Log pseudolikelihood = -304885.18

Pseudo R2 = 0.0638

1

race2 |

Black NH | -.1713457 .0416991 -4.11 0.000 -.2530745 -.0896169

Asian NH | .0389244 .0778322 0.50 0.617 -.113624 .1914728

NH/PI NH | -.7325147 .0657311 -11.14 0.000 -.8613453 -.6036841

AI/AN NH | .0881254 .0659335 1.34 0.181 -.041102 .2173528

Other NH | .0264418 .0678611 0.39 0.697 -.1065636 .1594472

Multiracial NH | .1890309 .0715008 2.64 0.008 .0488919 .3291698

Hispanic | -.7107596 .0582469 -12.20 0.000 -.8249214 -.5965977

|

income2 |

\$10k–14,999 | .0891214 .0179833 4.96 0.000 .0538748 .1243679

\$15k–19,999 | -.1964916 .0212999 -9.22 0.000 -.2382387 -.1547445

\$20k–24,999 | -.4533463 .0233161 -19.44 0.000 -.499045 -.4076476

\$25k–34,999 | -.7579439 .0219826 -34.48 0.000 -.801029 -.7148588

\$35k–49,999 | -.9606313 .0232003 -41.41 0.000 -1.006103 -.9151595

\$50k–74,999 | -1.184124 .0256463 -46.17 0.000 -1.23439 -1.133858

\$75k or more | -1.5003 .0245533 -61.10 0.000 -1.548424 -1.452177

|

educa |

Grades 1–8 | -.0905558 .107379 -0.84 0.399 -.3010148 .1199032

Some high school | -.2752582 .1037494 -2.65 0.008 -.4786034 -.0719131

High school graduate .. | -.5891852 .1041465 -5.66 0.000 -.7933086 -.3850619

Some college or tech .. | -.6669165 .0998275 -6.68 0.000 -.8625748 -.4712582

College graduate | -1.036101 .0980589 -10.57 0.000 -1.228293 -.8439092

|

\_cons | -1.420252 .1312281 -10.82 0.000 -1.677454 -1.163049

```

-----

.   eststo heart_attack

.   margins expansion_state#post2014

```

Predictive margins                      Number of obs = 2,274,241

Model VCE: Robust

Expression: Pr(heart\_attack), predict()

```

-----

      |      Delta-method
      |  Margin   std. err.   z   P>|z|   [95% conf. interval]
-----+-----
expansion_state#post2014 |
      0 0 | .0327448 .0011695  28.00  0.000   .0304526   .035037
      0 1 | .0331944 .0011282  29.42  0.000   .0309832   .0354056
      1 0 | .0313722 .0013786  22.76  0.000   .0286702   .0340742
      1 1 | .0323112 .0013909  23.23  0.000   .0295852   .0350373
-----

```

```

.   margins, dydx(post2014) at(expansion_state=(0 1))

```

Average marginal effects                      Number of obs = 2,274,241

Model VCE: Robust

Expression: Pr(heart\_attack), predict()

dy/dx wrt: 1.post2014

1.\_at: expansion\_state = 0

2.\_at: expansion\_state = 1

-----						
	Delta-method					
	dy/dx   std. err.   z   P> z    [95% conf. interval]					
-----+-----						
0.post2014	(base outcome)					
-----+-----						
1.post2014						
_at						
1		.0004496	.0005334	0.84	0.399	-.0005959 .0014951
2		.000939	.0004891	1.92	0.055	-.0000197 .0018977
-----						

Note: dy/dx for factor levels is the discrete change from the base level.

. \* stroke

.    logit stroke i.expansion\_state##i.post2014 ///    i.sex i.race2 i.income2 i.educa,  
vce(cluster \_state)

Iteration 0: Log pseudolikelihood = -255377.4

Iteration 1: Log pseudolikelihood = -247287.28

Iteration 2: Log pseudolikelihood = -238507.91

Iteration 3: Log pseudolikelihood = -238468.6



Iteration 4: Log pseudolikelihood = -238468.58

Iteration 5: Log pseudolikelihood = -238468.58

Logistic regression

Number of obs = 2,277,190

Wald chi2(25) = 27042.51

Prob > chi2 = 0.0000

Log pseudolikelihood = -238468.58

Pseudo R2 = 0.0662

(Std. err. adjusted for 54 clusters in \_state)

```
-----
|               Robust
stroke | Coefficient std. err.   z   P>|z|   [95% conf. interval]
-----+-----
1.expansion_state | -.0747959 .0534997  -1.40  0.162  -.1796535  .0300616
1.post2014 | .1400674 .0176159   7.95  0.000   .1055408  .1745939
|
expansion_state#post2014 |
1 1 | .0037871 .0256746   0.15  0.883  -.0465342  .0541085
|
sex |
Female | -.0677839 .0125045  -5.42  0.000  -.0922923  -.0432755
|
race2 |
Black NH | .2424962 .0395399   6.13  0.000   .1649994  .3199931
Asian NH | .0743683 .0674512   1.10  0.270  -.0578335  .2065702
NH/PI NH | -.6582596 .160191  -4.11  0.000  -.9722282  -.3442911
```

AI/AN NH	.1521988	.0714384	2.13	0.033	.0121822	.2922154
Other NH	.198707	.0589019	3.37	0.001	.0832613	.3141527
Multiracial NH	.3519492	.0653588	5.38	0.000	.2238484	.48005
Hispanic	-.729835	.0651069	-11.21	0.000	-.8574423	-.6022278
income2						
\$10k–14,999	.045282	.0216157	2.09	0.036	.002916	.0876479
\$15k–19,999	-.2789763	.0230385	-12.11	0.000	-.3241309	-.2338218
\$20k–24,999	-.5842095	.0276094	-21.16	0.000	-.638323	-.530096
\$25k–34,999	-.9388791	.0291868	-32.17	0.000	-.9960841	-.8816741
\$35k–49,999	-1.251612	.0304628	-41.09	0.000	-1.311317	-1.191906
\$50k–74,999	-1.542593	.0323263	-47.72	0.000	-1.605952	-1.479235
\$75k or more	-1.853606	.0337595	-54.91	0.000	-1.919773	-1.787438
educa						
Grades 1–8	-.0954619	.1175672	-0.81	0.417	-.3258894	.1349656
Some high school	-.1683056	.1113963	-1.51	0.131	-.3866382	.0500271
High school graduate ..	-.4431389	.1187426	-3.73	0.000	-.6758701	-.2104078
Some college or tech ..	-.4573151	.1145259	-3.99	0.000	-.6817816	-.2328485
College graduate	-.7103451	.1178462	-6.03	0.000	-.9413193	-.4793708
_cons	-2.204416	.1349225	-16.34	0.000	-2.468859	-1.939973

- 
- . eststo stroke
  - . margins expansion\_state#post2014

Predictive margins

Number of obs = 2,277,190

Model VCE: Robust

Expression: Pr(stroke), predict()

-----						
	Delta-method					
	Margin   std. err.   z   P> z    [95% conf. interval]					
-----+-----						
expansion_state#post2014						
0 0	.0228078	.0007837	29.10	0.000	.0212718	.0243437
0 1	.0260804	.0008506	30.66	0.000	.0244133	.0277476
1 0	.0212254	.0007835	27.09	0.000	.0196897	.0227611
1 1	.0243689	.0009348	26.07	0.000	.0225367	.0262011
-----						

.   margins, dydx(post2014) at(expansion\_state=(0 1))

Average marginal effects

Number of obs = 2,277,190

Model VCE: Robust

Expression: Pr(stroke), predict()

dy/dx wrt: 1.post2014

1.\_at: expansion\_state = 0

2.\_at: expansion\_state = 1

-----						
	Delta-method					
	dy/dx   std. err.   z   P> z    [95% conf. interval]					
-----+-----						
0.post2014	(base outcome)					
-----+-----						
1.post2014						
	_at					
1		.0032727	.0004144	7.90	0.000	.0024604 .0040849
2		.0031435	.0004293	7.32	0.000	.002302 .003985

-----

Note: dy/dx for factor levels is the discrete change from the base level.