

name: Mirella Charros

log: /Users/mirellacharros/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
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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.

-----+-----

0 | 3,026,198 73.30 73.30

1 | 1,102,152 26.70 100.00

-----+-----

Total | 4,128,350 100.00

. * 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]	
<hr/>						
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 -.118504
\$75k or more | -1.447293 .0267083 -54.19 0.000 -1.49964 -1.394945
|

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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

. * 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

\$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
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Some high school	.3805839	.0488702	7.79	0.000	.2848001	.4763678
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High school graduate ..	.1842981	.0490007	3.76	0.000	.0882585	.2803377
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Some college or tech ..	.2320091	.0504529	4.60	0.000	.1331234	.3308949
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College graduate	.0173351	.051197	0.34	0.735	-.0830091	.1176793
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|

_cons	1.757334	.0581292	30.23	0.000	1.643402	1.871265
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/lnalpha	1.859504	.0169191			1.826343	1.892665
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alpha	6.420553	.10863			6.211134	6.637033
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- eststo mentlhth

- 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]	
<hr/>						
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

. eststo has_doc

. margins expansion_state#post2014

Predictive margins

Number of obs = 2,320,108

Model VCE: Robust

Expression: Pr(has_doc), predict()

	Delta-method						
	Margin	std. err.	z	P> z	[95% conf. interval]		
<hr/>							
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(has_doc), predict()

dy/dx wrt: 1.post2014

1. at: expansion state = 0

2._at: expansion_state = 1

	dy/dx	std. err.	z	P> z	[95% conf. interval]
<hr/>					
0.post2014 (base outcome)					
<hr/>					

```
1.post2014 |  
    _at |  
1 | -.0225824 .0033788 -6.68 0.000 -.0292048 -.0159599  
2 | -.0125747 .0058827 -2.14 0.033 -.0241045 -.0010448
```

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

```
. * effect on medcost  
. generate couldnt_afford = .  
(2,684,074 missing values generated)  
. replace couldnt_afford = 1 if medcost == 1  
(397,162 real changes made)  
. replace couldnt_afford = 0 if medcost == 2  
(2,279,962 real changes made)  
. label variable couldnt_afford "Could not see doctor due to cost"  
. logit couldnt_afford i.expansion_state##i.post2014 ///  
> i.sex i.race2 i.income2 i.educa, vce(cluster _state)
```

Iteration 0: Log pseudolikelihood = -971889.17

Iteration 1: Log pseudolikelihood = -875627.76

Iteration 2: Log pseudolikelihood = -861870.56

Iteration 3: Log pseudolikelihood = -861761.22

Iteration 4: Log pseudolikelihood = -861761.18

Logistic regression Number of obs = 2,322,062

Wald chi2(25) = 18208.43

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
educa	
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()

	Delta-method						
	Margin	std. err.	z	P> z	[95% conf. interval]		
<hr/>							
expansion_state#post2014							
0 0	.1623159	.0058886	27.56	0.000	.1507745	.1738573	
0 1	.1499114	.0048793	30.72	0.000	.1403482	.1594746	
1 0	.1492877	.0059875	24.93	0.000	.1375525	.1610229	
1 1	.1268646	.0032376	39.18	0.000	.1205189	.1332102	

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

Average marginal effects Number of obs = 2,322,062

Model VCE: Robust

Expression: Pr(couldnt_afford), 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]		
<hr/>							
0.post2014 (base outcome)							
<hr/>							
1.post2014							

```
_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

10	.6820564	.0140421	48.57	0.000	.6545344	.7095785
11	.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]	
<hr/>						
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 0: Log pseudolikelihood = -325670.64

Iteration 1: Log pseudolikelihood = -310876.33

Iteration 2: Log pseudolikelihood = -304930.56

Iteration 3: Log pseudolikelihood = -304885.21

Iteration 4: Log pseudolikelihood = -304885.18

Iteration 5: Log pseudolikelihood = -304885.18

Logistic regression Number of obs = 2,274,241

Wald chi2(25) = 24729.31

Prob > chi2 = 0.0000

Log pseudolikelihood = -304885.18 Pseudo R2 = 0.0638

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

| Robust

heart_attack | Coefficient std. err. z P>|z| [95% conf. interval]

-----+-----
1.expansion_state | -.045249 .0606429 -0.75 0.456 -.1641068 .0736089

1.post2014 | .0144328 .0171652 0.84 0.400 -.0192104 .0480761

|

expansion_state#post2014 |

1 1 | .0167206 .0239114 0.70 0.484 -.0301449 .0635861

|

sex |

Female | -.731167 .0146546 -49.89 0.000 -.7598896 -.7024445

|

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]	
<hr/>						
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]	
<hr/>						
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]	
<hr/>						
0.post2014 (base outcome)						
<hr/>						
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.