

Advanced Methods in Health Services Research: Analysis - 309.716

Tuesday and Thursday 10:30-11:50

Instructors: Darrell J. Gaskin, Ph.D.

Roland J. Thorpe, Ph.D.

250 Hampton House

Computer Exercise #6: Estimating Multinomial Logit, Two Part Probit-OLS and GLM Models

Due: October 6, 2016

Use the analysis file built for computer exercise #1 and the subpop command to reduce the sample to **those persons 18 years to 64 years old**.

1. Estimate a multinomial logit model for the determinants of insurance status.

- a. Insurance status = $f(\text{health status, age, race, sex, poverty status, education, and location})$. Use the following reference categories: *excellent health, age category 25-44, male, white non-Hispanic, HS diploma/GED, high income, msa, East*).

```
. svy, subpop(adult): mlogit insurance ib3.agecat ib1.sex ib1.race ib5.income msa ib3.education
ib1.region ib1.healthstatus
(running mlogit on estimation sample)
```

Survey: Multinomial logistic regression

Number of strata	=	165	Number of obs	=	30158
Number of PSUs	=	370	Population size	=	269717779
			Subpop. no. of obs	=	15711
			Subpop. size	=	154522136
			Design df	=	205
			F(69, 137)	=	61.17
			Prob > F	=	0.0000

		Linearized					
insurance		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
Private_Insurance		(base outcome)					
Public_Insurance							
agecat							
18-24		.4052911	.1160632	3.49	0.001	.1764604	.6341217
45-64		-.4694487	.107731	-4.36	0.000	-.6818515	-.2570458
2.sex		.6562593	.0809203	8.11	0.000	.4967165	.815802
race							
Black		.7201112	.1205439	5.97	0.000	.4824463	.957776
Hispanic		.4557957	.1434915	3.18	0.002	.1728875	.738704
Asian		.1314794	.2329901	0.56	0.573	-.3278848	.5908436
Other		.0533022	.2197475	0.24	0.809	-.3799527	.4865571
income							
Poor		4.251985	.2223543	19.12	0.000	3.81359	4.690379
Near Poor		3.26231	.2316767	14.08	0.000	2.805536	3.719085
Low Income		2.659456	.206297	12.89	0.000	2.252721	3.066192
Middle Income		1.413408	.2123667	6.66	0.000	.9947056	1.832111
msa		-.0823923	.1717735	-0.48	0.632	-.4210614	.2562769
education							

No High School	.917141	.161076	5.69	0.000	.5995629	1.234719
Some High	.8500778	.1059918	8.02	0.000	.641104	1.059052
Some College/Tech School/AA degree	-.429354	.1033791	-4.15	0.000	-.6331765	-.2255315
College	-1.298107	.1671272	-7.77	0.000	-1.627615	-.9685984
region						
South	-1.419974	.1544701	-9.19	0.000	-1.724527	-1.11542
Midwest	-.7757658	.1772446	-4.38	0.000	-1.125222	-.4263097
West	-.6163193	.1912307	-3.22	0.001	-.9933504	-.2392881
healthstatus						
Very Good	.1389469	.1332466	1.04	0.298	-.1237626	.4016564
Good	.4486975	.1205004	3.72	0.000	.2111185	.6862764
Fair	.9456693	.1326335	7.13	0.000	.6841687	1.20717
Poor	2.090825	.1989954	10.51	0.000	1.698485	2.483165
_cons	-4.147342	.2881529	-14.39	0.000	-4.715466	-3.579219

Uninsured						
agecat						
18-24	-.0220343	.0744044	-0.30	0.767	-.1687302	.1246617
45-64	-.2315221	.0726922	-3.18	0.002	-.3748424	-.0882019
2.sex	-.3338661	.0531033	-6.29	0.000	-.4385648	-.2291674
race						
Black	.2302198	.0863242	2.67	0.008	.0600227	.4004169
Hispanic	.9295126	.0975924	9.52	0.000	.7370992	1.121926
Asian	.3393797	.1786468	1.90	0.059	-.0128409	.6916003
Other	.0276913	.1678672	0.16	0.869	-.3032762	.3586588
income						
Poor	2.493157	.1099588	22.67	0.000	2.276362	2.709953
Near Poor	1.884362	.1537194	12.26	0.000	1.581288	2.187436
Low Income	1.41562	.0997428	14.19	0.000	1.218967	1.612273
Middle Income	.6831492	.101556	6.73	0.000	.4829209	.8833774
msa	-.0302029	.0951153	-0.32	0.751	-.2177325	.1573267
education						
No High School	.8875245	.1601948	5.54	0.000	.5716839	1.203365
Some High	.431207	.0872598	4.94	0.000	.2591654	.6032486
Some College/Tech School/AA degree	-.3344949	.0842183	-3.97	0.000	-.50054	-.1684497
College	-1.021857	.1000081	-10.22	0.000	-1.219034	-.8246812
region						
South	.3637485	.1008367	3.61	0.000	.1649385	.5625585
Midwest	.0157004	.1100296	0.14	0.887	-.2012344	.2326351
West	.1326453	.1123637	1.18	0.239	-.0888914	.3541821
healthstatus						
Very Good	.0573467	.0889328	0.64	0.520	-.1179935	.2326869
Good	-.0623519	.1001388	-0.62	0.534	-.259786	.1350821
Fair	.0841687	.1299178	0.65	0.518	-.1719778	.3403152
Poor	.3409962	.2021567	1.69	0.093	-.0575767	.7395692
_cons	-2.111545	.1499035	-14.09	0.000	-2.407095	-1.815994

Medicare						
agecat						
18-24	-1.14137	.3885654	-2.94	0.004	-1.907467	-.3752733
45-64	.9796771	.1532432	6.39	0.000	.6775424	1.281812
2.sex	-.228388	.1198402	-1.91	0.058	-.4646654	.0078893
race						
Black	.0345603	.1494899	0.23	0.817	-.2601746	.3292951
Hispanic	-.6233952	.2088881	-2.98	0.003	-1.03524	-.2115507
Asian	-.4915361	.6965881	-0.71	0.481	-1.864932	.8818594
Other	.3143378	.4559502	0.69	0.491	-.5846153	1.213291

	income						
	Poor	3.384063	.2729184	12.40	0.000	2.845976	3.92215
	Near Poor	2.906522	.3461802	8.40	0.000	2.223992	3.589052
	Low Income	2.181762	.2702795	8.07	0.000	1.648878	2.714646
	Middle Income	1.15772	.2595938	4.46	0.000	.6459037	1.669536
	msa	-.0191817	.1703239	-0.11	0.910	-.3549928	.3166295
	education						
	No High School	.926049	.2247289	4.12	0.000	.4829727	1.369125
	Some High	.6553029	.1915686	3.42	0.001	.2776057	1.033
Some College/Tech School/AA degree		.0593331	.2150832	0.28	0.783	-.3647258	.4833919
College		-.7124037	.331956	-2.15	0.033	-1.366889	-.0579181
	region						
	South	-.4468301	.2129654	-2.10	0.037	-.8667133	-.0269469
	Midwest	-.6246119	.2266182	-2.76	0.006	-1.071413	-.1778106
	West	-.7360515	.293565	-2.51	0.013	-1.314845	-.1572578
	healthstatus						
	Very Good	.4630448	.3415361	1.36	0.177	-.2103289	1.136419
	Good	1.213139	.3052068	3.97	0.000	.6113925	1.814886
	Fair	2.103954	.3042284	6.92	0.000	1.504136	2.703772
	Poor	3.146354	.3464028	9.08	0.000	2.463385	3.829323
	_cons	-5.761939	.4446996	-12.96	0.000	-6.63871	-4.885168

b. Compute and interpret the relative rate ratios for poverty categories. Compute and interpret the marginal effects of being poor and near poor on being uninsured or publicly insured.

```
. svy, subpop(adult): mlogit insurance ib3.agecat ib1.sex ib1.race ib5.income msa ib3.education
ib1.region ib1.healthstatus, rrr
(running mlogit on estimation sample)
```

Survey: Multinomial logistic regression

Number of strata	=	165	Number of obs	=	30158
Number of PSUs	=	370	Population size	=	269717779
			Subpop. no. of obs	=	15711
			Subpop. size	=	154522136
			Design df	=	205
			F(69, 137)	=	61.17
			Prob > F	=	0.0000

		Linearized					
insurance		RRR	Std. Err.	t	P> t	[95% Conf. Interval]	
Private_Insurance		(base outcome)					
Public_Insurance							
	agecat						
	18-24	1.499739	.1740645	3.49	0.001	1.192987	1.885365
	45-64	.6253469	.0673693	-4.36	0.000	.5056798	.7733328
	2.sex	1.927568	.1559794	8.11	0.000	1.643317	2.260988
	race						
	Black	2.054662	.247677	5.97	0.000	1.620033	2.605895
	Hispanic	1.577428	.2263474	3.18	0.002	1.188732	2.093221
	Asian	1.140514	.2657286	0.56	0.573	.720446	1.805511
	Other	1.054748	.2317783	0.24	0.809	.6838938	1.626706
	income						
	Poor	70.2447	15.61921	19.12	0.000	45.31284	108.8945

	Near Poor	26.10979	6.04903	14.08	0.000	16.53593	41.22665
	Low Income	14.28852	2.947678	12.89	0.000	9.513583	21.46003
	Middle Income	4.10994	.8728143	6.66	0.000	2.703928	6.247062
	msa	.9209107	.158188	-0.48	0.632	.6563498	1.29211
	education						
	No High School	2.502127	.4030326	5.69	0.000	1.821323	3.437413
	Some High	2.339829	.2480027	8.02	0.000	1.898576	2.883635
Some College/Tech School/AA degree		.6509295	.0672925	-4.15	0.000	.5309027	.798092
	College	.2730482	.0456338	-7.77	0.000	.1963974	.3796147
	region						
	South	.2417204	.0373386	-9.19	0.000	.1782573	.3277776
	Midwest	.4603511	.0815948	-4.38	0.000	.3245804	.6529141
	West	.5399281	.1032509	-3.22	0.001	.3703338	.7871881
	healthstatus						
	Very Good	1.149063	.1531088	1.04	0.298	.8835895	1.494298
	Good	1.566271	.1887362	3.72	0.000	1.235059	1.986306
	Fair	2.574536	.3414697	7.13	0.000	1.982123	3.344008
	Poor	8.091586	1.610188	10.51	0.000	5.465659	11.97911
	_cons	.0158064	.0045547	-14.39	0.000	.0089557	.0278975

Uninsured	agecat						
	18-24	.9782067	.0727829	-0.30	0.767	.8447368	1.132765
	45-64	.7933251	.0576686	-3.18	0.002	.6873976	.915576
	2.sex	.7161497	.0380299	-6.29	0.000	.6449614	.7951954
	race						
	Black	1.258877	.1086715	2.67	0.008	1.061861	1.492447
	Hispanic	2.533274	.2472282	9.52	0.000	2.089864	3.070763
	Asian	1.404076	.2508337	1.90	0.059	.9872412	1.996909
	Other	1.028078	.1725806	0.16	0.869	.7383951	1.431408
	income						
	Poor	12.09942	1.330438	22.67	0.000	9.74118	15.02856
	Near Poor	6.582153	1.011805	12.26	0.000	4.861214	8.912329
	Low Income	4.119039	.4108444	14.19	0.000	3.38369	5.014196
	Middle Income	1.980104	.2010915	6.73	0.000	1.620802	2.419056
	msa	.9702487	.0922855	-0.32	0.751	.8043406	1.170378
	education						
	No High School	2.429109	.3891306	5.54	0.000	1.771247	3.331308
	Some High	1.539114	.1343027	4.94	0.000	1.295848	1.828048
Some College/Tech School/AA degree		.7156995	.060275	-3.97	0.000	.6062032	.8449737
	College	.3599258	.0359955	-10.22	0.000	.2955156	.4383747
	region						
	South	1.438712	.145075	3.61	0.000	1.179321	1.755157
	Midwest	1.015824	.1117707	0.14	0.887	.8177208	1.261921
	West	1.141845	.128302	1.18	0.239	.9149449	1.425015
	healthstatus						
	Very Good	1.059023	.0941819	0.64	0.520	.8887018	1.261986
	Good	.9395522	.0940856	-0.62	0.534	.7712166	1.144631
	Fair	1.087812	.1413262	0.65	0.518	.8419979	1.40539
	Poor	1.406348	.2843027	1.69	0.093	.9440495	2.095033
	_cons	.1210508	.0181459	-14.09	0.000	.0900766	.1626761

Medicare	agecat						
	18-24	.3193811	.1241004	-2.94	0.004	.1484559	.6871015
	45-64	2.663596	.4081779	6.39	0.000	1.969033	3.603162
	2.sex	.7958154	.0953707	-1.91	0.058	.6283453	1.00792

	race						
	Black	1.035164	.1547466	0.23	0.817	.770917	1.389988
	Hispanic	.5361211	.1119893	-2.98	0.003	.3551413	.8093282
	Asian	.611686	.4260932	-0.71	0.481	.1549068	2.415387
	Other	1.369352	.6243565	0.69	0.491	.5573202	3.364539
	income						
	Poor	29.49035	8.048458	12.40	0.000	17.21836	50.50892
	Near Poor	18.29306	6.332695	8.40	0.000	9.244157	36.19973
	Low Income	8.861905	2.395191	8.07	0.000	5.20114	15.09926
	Middle Income	3.182668	.826201	4.46	0.000	1.90771	5.309703
	msa	.9810011	.1670879	-0.11	0.910	.7011785	1.372494
	education						
	No High School	2.524515	.5673316	4.12	0.000	1.620886	3.93191
	Some High	1.925726	.3689085	3.42	0.001	1.319966	2.809482
Some College/Tech School/AA degree		1.061129	.228231	0.28	0.783	.694387	1.621565
College		.4904638	.1628124	-2.15	0.033	.2548986	.9437273
	region						
	South	.6396526	.1362238	-2.10	0.037	.4203308	.973413
	Midwest	.5354692	.1213471	-2.76	0.006	.3425241	.8371009
	West	.4790015	.1406181	-2.51	0.013	.2685159	.8544837
	healthstatus						
	Very Good	1.588905	.5426683	1.36	0.177	.8103177	3.11559
	Good	3.364029	1.026724	3.97	0.000	1.842996	6.140377
	Fair	8.198524	2.494223	6.92	0.000	4.500266	14.93596
	Poor	23.25113	8.054257	9.08	0.000	11.7445	46.03135
	_cons	.003145	.0013986	-12.96	0.000	.0013087	.0075579

The relative risk ratio describe the relative likelihood of belonging in the respective categories, i.e. public insurance, uninsured, Medicare, compared to having private insurance when you fall into one of the poverty categories. For example, being poor increase the relative risk that one would be uninsured to 12.10 compared to having private insurance. Similarly, being poor increases the relative risk that one would have Medicare to 29.49 compared to having private insurance. The other coefficients can have the same interpretation.

MARGINAL EFFECTS:

```
. margins, dydx(1.income 2.income) predict(outcome(2)) noestimcheck /*publicly insured not medicare*/
```

```
Average marginal effects          Number of obs   =       25152
Model VCE      : Linearized
```

```
Expression      : Pr(insurance==Public_Insurance), predict(outcome(2))
dy/dx w.r.t.    : 1.income 2.income
```

		Delta-method				
		dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]

income						
	Poor	.2455239	.0138132	17.77	0.000	.2184506 .2725973
	Near Poor	.1527979	.0154428	9.89	0.000	.1225305 .1830653

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

```
. margins, dydx(1.income 2.income) predict(outcome(3)) noestimcheck /*uninsured*/
```

```
Average marginal effects      Number of obs   =      25152
Model VCE      : Linearized
```

```
Expression      : Pr(insurance==Uninsured), predict(outcome(3))
dy/dx w.r.t.    : 1.income 2.income
```

		Delta-method					
		dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]	
income							
Poor		.2684913	.0179669	14.94	0.000	.2332769	.3037058
Near Poor		.224161	.0255516	8.77	0.000	.1740808	.2742412

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

The marginal effects measure the direct effect of being poor or near poor on the log odds of being publicly insured or uninsured. Specifically, being poor increases the log odds of being publicly insured by 0.246 while being near poor increases the log odds of being publicly insured by 0.153. Similarly, being poor increases the log odds of being uninsured by 0.268 while being near poor increases the log odds of being uninsured by 0.224.

2. For persons 18-64, estimate the following models of health expenditures. Use the following specification: $\text{Log (Total health expenditures)} = f(\text{health status, age, race, gender, education, insurance status, poverty status and location})$. *Use the following reference categories: excellent health, age category 25-44, male, white non-Hispanic, HS diploma/GED, privately insured, high income, msa, East*
 - a. Estimate the two part probit-OLS model. Create a table displaying the coefficients and marginal effects of income and education on health expenditures.

	First Stage (Probit)	2nd Stage (OLS)
income		
Poor	-0.285*** (0.0526)	-0.150** (0.0600)
Near Poor	-0.298*** (0.0698)	-0.235*** (0.0859)
Low Income	-0.179*** (0.0559)	-0.189*** (0.0604)
Middle Income	-0.201*** (0.0401)	-0.0934** (0.0389)
education		
Grade 8 and below	-0.0293 (0.0642)	-0.177** (0.0777)
Some High School	-0.0381 (0.0491)	-0.133** (0.0579)

Some College	0.205***	0.0924**
	(0.0404)	(0.0405)
College	0.323***	0.180***
	(0.0475)	(0.0465)
Advanced Degree	0.336***	0.282***
	(0.0683)	(0.0585)

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

b. Conduct a Park Test to determine what GLM analysis to estimate.

```
. /*Part B*/
.
. /*Park Test*/
. svy, subpop(age18to64): glm healthexp ib3.agecat ib1.sex ib1.race ib5.income notmsa ib3.education
ib1.region ib1.healthstatus ib
> 1.insurance, family(gamma) link(log)
(running glm on estimation sample)
```

Survey: Generalized linear models

Number of strata	=	165	Number of obs	=	31113
Number of PSUs	=	370	Population size	=	282464364
			Subpop. no. of obs	=	16917
			Subpop. size	=	170277348
			Design df	=	205

healthexp		Linearized					
		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
agecat							
18-24		-.1525881	.0708288	-2.15	0.032	-.2922343	-.0129418
45-64		.4421097	.0433598	10.20	0.000	.3566214	.527598
sex							
female		.4685999	.0384777	12.18	0.000	.392737	.5444627
race							
Black		-.4413024	.0592257	-7.45	0.000	-.5580719	-.3245329
Hispanic		-.5400887	.0616302	-8.76	0.000	-.661599	-.4185783
Asian		-.7675889	.0722038	-10.63	0.000	-.9099462	-.6252317
Other		.0192808	.1349877	0.14	0.887	-.2468613	.285423
income							
Poor		-.1192354	.088956	-1.34	0.182	-.2946214	.0561505
Near Poor		-.1538654	.093684	-1.64	0.102	-.3385731	.0308423
Low Income		-.2147216	.0717328	-2.99	0.003	-.3561503	-.073293
Middle Income		-.1369234	.0433445	-3.16	0.002	-.2223815	-.0514653
notmsa		-.1861548	.0524585	-3.55	0.000	-.2895822	-.0827274
education							
Grade 8 and below		-.0523859	.1500581	-0.35	0.727	-.3482409	.2434692
Some High School		-.0763148	.0736007	-1.04	0.301	-.2214262	.0687966
Some College		.1285256	.0481397	2.67	0.008	.0336131	.2234381
College		.2032447	.0560938	3.62	0.000	.0926499	.3138394
Advanced Degree		.2645275	.0751969	3.52	0.001	.1162689	.412786
region							
Midwest		.0598253	.0560905	1.07	0.287	-.050763	.1704136
South		.0173792	.0572827	0.30	0.762	-.0955596	.130318
West		-.0469751	.0534701	-0.88	0.381	-.152397	.0584468

healthstatus							
Very Good		.3019975	.0573575	5.27	0.000	.1889112	.4150838
Good		.6460287	.0520156	12.42	0.000	.5434745	.7485829
Fair		1.192653	.0652875	18.27	0.000	1.063932	1.321374
Poor		1.740819	.0787071	22.12	0.000	1.58564	1.895998
insurance							
Public Ins not Medicare		.3835083	.0833456	4.60	0.000	.2191839	.5478327
Uninsured		-.9190729	.0725297	-12.67	0.000	-1.062073	-.7760731
Public Ins Medicare		.7460566	.0923205	8.08	0.000	.5640372	.928076
_cons		7.157574	.0815598	87.76	0.000	6.996771	7.318378

```
. predict healthhat2, xb
(6345 missing values generated)
```

```
. gen lnhealthhat2 = ln(healthhat2)
(6345 missing values generated)
```

```
. gen r2 = ((healthexp-healthhat2))^2
(6526 missing values generated)
```

```
. gen lnr2 = ln(r2)
(6526 missing values generated)
```

```
. svy, subpop(age18to64): glm lnr2 lnhealthhat2, family(gamma) link(log) nolog
(running glm on estimation sample)
```

Survey: Generalized linear models

Number of strata	=	165	Number of obs	=	31113
Number of PSUs	=	370	Population size	=	282464364
			Subpop. no. of obs	=	16917
			Subpop. size	=	170277348
			Design df	=	205

		Linearized				
lnr2		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnhealthhat2		2.230204	.0444583	50.16	0.000	2.142549 2.317858
_cons		-2.048806	.0920827	-22.25	0.000	-2.230357 -1.867256

```
. test lnhealthhat2 == 0
```

Adjusted Wald test

```
( 1) [lnr2]lnhealthhat2 = 0
      F( 1, 205) = 2516.42
      Prob > F = 0.0000
```

```
. test lnhealthhat2 == 1
```

Adjusted Wald test

```
( 1) [lnr2]lnhealthhat2 = 1
      F( 1, 205) = 765.68
      Prob > F = 0.0000
```

```
. test lnhealthhat2 == 2
```

Adjusted Wald test

```
( 1) [lnr2]lnhealthhat2 = 2
      F( 1, 205) = 26.81
      Prob > F = 0.0000
```



```
. test lnhealthhat2 == 3

Adjusted Wald test

( 1)  [lnr2]lnhealthhat2 = 3

      F( 1, 205) = 299.81
      Prob > F = 0.0000
```

We see that the coefficient is 2.2. It is nearest to 2 and thus we use the Gamma distribution.

- c. Estimate the GLM (one step). Create a table display the coefficients and marginal effects of poverty status

	GLM	Marginal Effects
income		
Poor	-0.119	-363.89
	(0.0890)	(263.60)
Near Poor	-0.154	-461.70*
	(0.0937)	(268.46)
Low Income	-0.215***	-625.58***
	(0.0717)	(202.75)
Middle Income	-0.137***	-414.27***
	(0.0433)	(132.38)

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

- d. Compare the GLM result to the two part probit-OLS results.

	First Stage (Probit)	2nd Stage (OLS)	GLM
income			
Poor	-0.285***	-0.150**	-0.119
	(0.0526)	(0.0600)	(0.0890)
Near Poor	-0.298***	-0.235***	-0.154
	(0.0698)	(0.0859)	(0.0937)
Low Income	-0.179***	-0.189***	-0.215***
	(0.0559)	(0.0604)	(0.0717)
Middle Income	-0.201***	-0.0934**	-0.137***
	(0.0401)	(0.0389)	(0.0433)

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

In the two part probit-OLS model, we see that each poverty status category is statistically significant in decreasing the probability of having any expenditures. In addition, they also significantly decrease health expenditures by 15% (poor), 23.5% (near poor), 18.9% (low income), and 9.34% (high income), conditional on having any health expenditures. On the other hand, relative to high-income individuals, only low-income and middle-income individuals significantly decreases expenditures by 21.5% and 13.7%, respectively.

- e. Compare the predict mean, 25th and 75th percentile from the GLM and two part probit-OLS models. (Note use a smearing factor to compute the predicted value for the two part probit-OLS model)

```
. gen predictolsmodel = smear*healthexphat*probithealthexpnhat
(6345 missing values generated)
```

```
. summarize predictolsmodel, detail
```

predictolsmodel					
Percentiles			Smallest		
1%	440.6569		200.7225		
5%	1134.039		200.7225		
10%	2200.526		221.8453	Obs	26721
25%	4875.182		225.2032	Sum of Wgt.	26721
50%	10272.64			Mean	18453.2
			Largest	Std. Dev.	27925.07
75%	21501.58		477914.6	Variance	7.80e+08
90%	40542.86		479207.3	Skewness	5.665885
95%	58902.71		482385.7	Kurtosis	56.15866
99%	134717.8		526405.1		

```
. quietly svy, subpop(ages18to64): glm healthexp ib3.agecat ib1.sex
ib1.race ib5.income notmsa ib3.education ib1.region ib1.healths
> tatus ib1.insurance, family(gamma) link(log)
```

```
. predict glmexppredict
(option mu assumed; predicted mean healthexp)
(6345 missing values generated)
```

```
. summarize glmexppredict, detail
```

Predicted mean healthexp					
Percentiles			Smallest		
1%	251.2586		155.1926		
5%	422.3792		157.2931		
10%	607.9389		161.7436	Obs	26721
25%	1076.563		170.8991	Sum of Wgt.	26721
50%	1828.039			Mean	2625.364
			Largest	Std. Dev.	2796.034
75%	3158.811		39124.57		

90%	5215.802	41121.04	Variance	7817808
95%	7386.246	41121.04	Skewness	4.001164
99%	14104.01	42803.04	Kurtosis	29.73515

.

We see that in the two part probit-OLS model, there is a wider spread of expenditures and we observe that the magnitudes of expenditures at the 25th and 75th percentiles are larger than those we see from the GLM model. The mean for the two part model is also much larger than that we see in the GLM model.

Log file

```
. do "C:\Users\bdela\AppData\Local\Temp\STD00000000.tmp"

. /*Variable Creation*/
.
.
. gen age18over = 0

. replace age18over = 1 if age08x >= 18
(23183 real changes made)

.
. gen agecat = 1 if age08x >=0 & age08x <=17
(23434 missing values generated)

. replace agecat = 2 if age08x >=18 & age08x <=24
(3372 real changes made)

. replace agecat = 3 if age08x >=25 & age08x <=44
(8813 real changes made)

. replace agecat = 4 if age08x >=45 & age08x <=64
(7614 real changes made)

. replace agecat = 5 if age08x >=65 & age08x <=74
(1867 real changes made)

. replace agecat = 6 if age08x >=75
(1517 real changes made)

.
. label define agecats 1 "0-18" 2 "18-24" 3 "25-44" 4 "45-64" 5 "65-74" 6 "75+"

. label values agecat agecats /*set label name sexn to the variable sex*/

. tabulate agecat
```

agecat	Freq.	Percent	Cum.
0-18	9,632	29.35	29.35
18-24	3,372	10.28	39.63
25-44	8,813	26.86	66.48
45-64	7,614	23.20	89.69
65-74	1,867	5.69	95.38
75+	1,517	4.62	100.00
Total	32,815	100.00	

```
.
.
.
. gen race = 1 if racex == 1 & racethnx != 1
(18775 missing values generated)

. replace race = 2 if racex == 2 & racethnx != 1
(6476 real changes made)

. replace race = 3 if racethnx == 1
(9392 real changes made)

. replace race = 4 if racex == 4 & racethnx != 1
(1997 real changes made)
```

```
. replace race = 5 if (racex == 3 | racex == 5 | racex == 6) & racethnx != 1
(910 real changes made)
```

```
. label define racexn2 1 "White" 2 "Black" 3 "Hispanic" 4 "Asian" 5 "Other"
```

```
. label values race racexn2
```

```
. label define sexn 1 "male" 2 "female"
```

```
. label values sex sexn
```

```
. tabulate sex
```

sex	Freq.	Percent	Cum.
-----+-----			
male	15,885	48.04	48.04
female	17,181	51.96	100.00
-----+-----			
Total	33,066	100.00	

```
. gen female = 0 if sex == 1
(17181 missing values generated)
```

```
. replace female = 1 if sex == 2
(17181 real changes made)
```

```
.
.
.
```

```
. gen racesexcat = 1 if race == 1 & female == 0
(26095 missing values generated)
```

```
. replace racesexcat = 2 if race == 1 & female == 1
(7320 real changes made)
```

```
. replace racesexcat = 3 if race == 2 & female == 0
(2894 real changes made)
```

```
. replace racesexcat = 4 if race == 2 & female == 1
(3582 real changes made)
```

```
. replace racesexcat = 5 if race == 3 & female == 0
(4612 real changes made)
```

```
. replace racesexcat = 6 if race == 3 & female == 1
(4780 real changes made)
```

```
. replace racesexcat = 7 if race == 4 & female == 0
(965 real changes made)
```

```
. replace racesexcat = 8 if race == 4 & female == 1
(1032 real changes made)
```

```
. replace racesexcat = 9 if race == 5 & female == 0
(443 real changes made)
```

```
. replace racesexcat = 10 if race == 5 & female == 1
(467 real changes made)
```

```

.
.
. gen whitemale = 0 /*okay since no missings for race and sex*/

. replace whitemale = 1 if race == 1 & female == 0
(6971 real changes made)

. gen whitefemale = 0

. replace whitefemale = 1 if race == 1 & female == 1
(7320 real changes made)

. gen blackmale = 0

. replace blackmale = 1 if race == 2 & female == 0
(2894 real changes made)

. gen blackfemale = 0

. replace blackfemale = 1 if race == 2 & female == 1
(3582 real changes made)

. gen hispanicmale = 0

. replace hispanicmale = 1 if race == 3 & female == 0
(4612 real changes made)

. gen hispanicfemale = 0

. replace hispanicfemale = 1 if race == 3 & female == 1
(4780 real changes made)

. gen asianmale = 0

. replace asianmale = 1 if race == 4 & female == 0
(965 real changes made)

. gen asianfemale = 0

. replace asianfemale = 1 if race == 4 & female == 1
(1032 real changes made)

. gen othermale = 0

. replace othermale = 1 if race == 5 & female == 0
(443 real changes made)

. gen otherfemale = 0

. replace otherfemale = 1 if race == 5 & female == 1
(467 real changes made)

.
. label define racesex 1 "White Male" 2 "White Female" 3 "Black Male" 4 "Black Female" 5
"Hispanic Male" 6 "Hispanic Female" 7 "Asian Male" 8 "Asian Female" 9 "Other Male" 10 "Other Female"
>
. label values racesexcat racesex

.
.
. gen healthstatus = 1 if rthlth42 == 1
(23215 missing values generated)

```

```

. replace healthstatus = 2 if rthlth42 == 2
(10188 real changes made)

. replace healthstatus = 3 if rthlth42 == 3
(8628 real changes made)

. replace healthstatus = 4 if rthlth42 == 4
(2777 real changes made)

. replace healthstatus = 5 if rthlth42 == 5
(815 real changes made)

.
. label define health 1 "Excellent" 2 "Very Good" 3 "Good" 4 "Fair" 5 "Poor"

. label values healthstatus health

.
.
.
. gen education = 1 if educyr >=0 & educyr <=8
(25465 missing values generated)

. replace education = 2 if educyr >=9 & (educyr <=12 & educyr >=0) & hideg == 1
(3732 real changes made)

. replace education = 3 if hideg == 2 | hideg == 3
(11173 real changes made)

. replace education = 4 if educyr >=13 & educyr <=17 & hideg == 3
(3629 real changes made)

. replace education = 5 if hideg == 4
(3245 real changes made)

. replace education = 6 if hideg == 5 | hideg ==6
(1590 real changes made)

.
. label define educ 1 "Grade 8 and below" 2 "Some High School" 3 "HighSchool" 4 "Some College"
5 "College" 6 "Advanced Degree"

. label values education educ

.
.
.
. gen insurance = 1 if inscov08 == 1
(14773 missing values generated)

. replace insurance = 2 if inscov08 == 2 & mcrev08 == 2
(6642 real changes made)

. replace insurance = 3 if inscov08 == 3
(5662 real changes made)

. replace insurance = 4 if inscov08 == 2 & mcrev08 == 1
(2469 real changes made)

.
. label define insure 1 "Private Ins" 2 "Public Ins not Medicare" 3 "Uninsured" 4 "Public Ins
Medicare"

```

```

. label values insurance insure

.
.
.
. gen region = 1 if region08 == 1
(28080 missing values generated)

. replace region = 2 if region08 == 2
(6499 real changes made)

. replace region = 3 if region08 == 3
(12424 real changes made)

. replace region = 4 if region08 == 4
(8906 real changes made)

.
. label define region2 1 "Northeast" 2 "Midwest" 3 "South" 4 "West"

. label values region region2

.
.
.
. gen income = 1 if povcat08 == 1
(26099 missing values generated)

. replace income = 2 if povcat08 == 2
(2171 real changes made)

. replace income = 3 if povcat08 == 3
(5667 real changes made)

. replace income = 4 if povcat08 == 4
(9595 real changes made)

. replace income = 5 if povcat08 == 5
(8666 real changes made)

.
. label define pov 1 "Poor" 2 "Near Poor" 3 "Low Income" 4 "Middle Income" 5 "High Income"

. label values income pov

.
. gen notmsa = 1 if msa08 == 0
(28406 missing values generated)

. replace notmsa = 0 if msa08 == 1
(28155 real changes made)

.
. gen obtotv08n = 0 if obtotv08 == 0
(22054 missing values generated)

. replace obtotv08n = 1 if obtotv08 >=1
(22054 real changes made)

.
.
.

```


Number of strata	=	165	Number of obs	=	31216
Number of PSUs	=	370	Population size	=	283647757
			Subpop. no. of obs	=	17020

Subpop. size = 171460741
Design df = 205
F(72, 134) = 57.75
Prob > F = 0.0000

		Linearized					
insurance		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
Private_Ins		(base outcome)					
Public_Ins_not_Medicare							
	agecat						
	18-24	.4250573	.1156715	3.67	0.000	.1969989	.6531156
	45-64	-.4344212	.1056984	-4.11	0.000	-.6428166	-.2260258
	sex						
	female	.6405807	.0793574	8.07	0.000	.4841195	.7970419
	race						
	Black	.7405813	.120224	6.16	0.000	.5035473	.9776153
	Hispanic	.4819201	.1432748	3.36	0.001	.1994391	.7644012
	Asian	.2161596	.2330734	0.93	0.355	-.2433687	.6756879
	Other	.0723553	.2197977	0.33	0.742	-.3609986	.5057092
	income						
	Poor	4.27075	.2204577	19.37	0.000	3.836095	4.705405
	Near Poor	3.295032	.2303908	14.30	0.000	2.840793	3.749272
	Low Income	2.671577	.2020799	13.22	0.000	2.273155	3.069998
	Middle Income	1.427483	.2092371	6.82	0.000	1.01495	1.840016
	notmsa	.0737096	.1714215	0.43	0.668	-.2642657	.4116849
	education						
	Grade 8 and below	.8997485	.1610539	5.59	0.000	.582214	1.217283
	Some High School	.8452103	.106203	7.96	0.000	.63582	1.0546
	Some College	-.4271726	.1035657	-4.12	0.000	-.6313631	-.222982
	College	-1.297537	.167468	-7.75	0.000	-1.627718	-.9673567
	Advanced Degree	-1.919394	.3961337	-4.85	0.000	-2.700413	-1.138376
	region						
	Midwest	-.7732221	.1778333	-4.35	0.000	-1.123839	-.4226054
	South	-1.419507	.1557189	-9.12	0.000	-1.726523	-1.112491
	West	-.6141853	.191275	-3.21	0.002	-.9913038	-.2370668
	healthstatus						
	Very Good	.1366326	.1321508	1.03	0.302	-.1239164	.3971816
	Good	.470287	.1180262	3.98	0.000	.2375862	.7029879
	Fair	.965964	.1309581	7.38	0.000	.7077664	1.224162
	Poor	2.120388	.2010133	10.55	0.000	1.724069	2.516706
	_cons	-4.269223	.2475735	-17.24	0.000	-4.75734	-3.781106
Uninsured							
	agecat						
	18-24	.0122333	.0742184	0.16	0.869	-.1340958	.1585625
	45-64	-.1799719	.0718844	-2.50	0.013	-.3216994	-.0382444
	sex						
	female	-.3404412	.0521949	-6.52	0.000	-.4433488	-.2375335
	race						
	Black	.255403	.083602	3.05	0.003	.0905731	.420233

Hispanic		.9495525	.0974765	9.74	0.000	.7573676	1.141737
Asian		.3285284	.1705771	1.93	0.055	-.007782	.6648388
Other		.026392	.1686852	0.16	0.876	-.3061883	.3589723
income							
Poor		2.50168	.1096096	22.82	0.000	2.285574	2.717787
Near Poor		1.899079	.1528869	12.42	0.000	1.597647	2.200511
Low Income		1.414986	.0986358	14.35	0.000	1.220515	1.609457
Middle Income		.6795717	.0985951	6.89	0.000	.4851812	.8739622
notmsa		.0132486	.0936087	0.14	0.888	-.1713106	.1978079
education							
Grade 8 and below		.8695881	.16205	5.37	0.000	.5500897	1.189086
Some High School		.4251489	.0875071	4.86	0.000	.2526196	.5976782
Some College		-.3317883	.08423	-3.94	0.000	-.4978565	-.1657202
College		-1.010168	.1000383	-10.10	0.000	-1.207404	-.8129322
Advanced Degree		-1.421426	.1689722	-8.41	0.000	-1.754572	-1.08828
region							
Midwest		.029095	.1138157	0.26	0.798	-.1953045	.2534945
South		.3648931	.1038454	3.51	0.001	.1601511	.5696351
West		.1544587	.1163394	1.33	0.186	-.0749165	.3838339
healthstatus							
Very Good		.0727436	.0876178	0.83	0.407	-.100004	.2454912
Good		-.0214427	.0995895	-0.22	0.830	-.2177937	.1749083
Fair		.1119966	.1289173	0.87	0.386	-.1421772	.3661703
Poor		.4366891	.2046015	2.13	0.034	.0332962	.8400821
_cons		-2.198516	.1454991	-15.11	0.000	-2.485383	-1.91165

Public_Ins_Medicare							
agecat							
18-24		-1.116939	.3889281	-2.87	0.005	-1.883751	-.3501265
45-64		1.025898	.1537105	6.67	0.000	.7228421	1.328954
sex							
female		-.2527068	.1179321	-2.14	0.033	-.4852222	-.0201915
race							
Black		.0605307	.1492601	0.41	0.686	-.233751	.3548123
Hispanic		-.6190991	.2081962	-2.97	0.003	-1.02958	-.2086186
Asian		-.5409819	.6919991	-0.78	0.435	-1.90533	.823366
Other		.4182817	.4388933	0.95	0.342	-.4470419	1.283605
income							
Poor		3.461414	.2788149	12.41	0.000	2.911702	4.011127
Near Poor		2.968715	.3501395	8.48	0.000	2.278379	3.659051
Low Income		2.242257	.2774753	8.08	0.000	1.695186	2.789328
Middle Income		1.209711	.2622096	4.61	0.000	.6927379	1.726685
notmsa		.0065116	.1693585	0.04	0.969	-.3273962	.3404193
education							
Grade 8 and below		.916701	.2253527	4.07	0.000	.4723949	1.361007
Some High School		.6524403	.1912873	3.41	0.001	.2752976	1.029583
Some College		.0598757	.2154081	0.28	0.781	-.3648237	.484575
College		-.705655	.3328669	-2.12	0.035	-1.361937	-.0493735
Advanced Degree		-1.536587	.5786653	-2.66	0.009	-2.677486	-.3956884
region							
Midwest		-.6156773	.2295321	-2.68	0.008	-1.068224	-.163131

South		-.4450503	.2151248	-2.07	0.040	-.8691911	-.0209095
West		-.6730808	.2904714	-2.32	0.021	-1.245775	-.1003864
healthstatus							
Very Good		.5196919	.3406673	1.53	0.129	-.151969	1.191353
Good		1.255214	.3065598	4.09	0.000	.6508	1.859629
Fair		2.150626	.3048072	7.06	0.000	1.549667	2.751585
Poor		3.185727	.3496673	9.11	0.000	2.496322	3.875132
_cons		-5.90819	.4575703	-12.91	0.000	-6.810337	-5.006043

```

. svy, subpop(age18to64): mlogit insurance ib3.agecat ib1.sex ib1.race ib5.income notmsa
ib3.education ib1.region ib1.healthstatus
> , rrr
(running mlogit on estimation sample)

```

Survey: Multinomial logistic regression

Number of strata	=	165	Number of obs	=	31216
Number of PSUs	=	370	Population size	=	283647757
			Subpop. no. of obs	=	17020
			Subpop. size	=	171460741
			Design df	=	205
			F(72, 134)	=	57.75
			Prob > F	=	0.0000

insurance		RRR	Linearized Std. Err.	t	P> t	[95% Conf. Interval]
Private_Ins		(base outcome)				
Public_Ins_not_Medicare						
agecat						
18-24		1.529678	.1769402	3.67	0.000	1.217743 1.921518
45-64		.6476394	.0684545	-4.11	0.000	.5258094 .7976975
sex						
female		1.897582	.1505871	8.07	0.000	1.622745 2.218967
race						
Black		2.097154	.2521282	6.16	0.000	1.65458 2.65811
Hispanic		1.61918	.2319877	3.36	0.001	1.220718 2.147708
Asian		1.241301	.2893141	0.93	0.355	.7839824 1.965385
Other		1.075037	.2362907	0.33	0.742	.69698 1.658161
income						
Poor		71.57527	15.77932	19.37	0.000	46.34412 110.543
Near Poor		26.97828	6.21555	14.30	0.000	17.12934 42.49012
Low Income		14.46276	2.922633	13.22	0.000	9.709992 21.54187
Middle Income		4.168195	.8721409	6.82	0.000	2.759227 6.296636
notmsa		1.076494	.1845343	0.43	0.668	.7677695 1.509359
education						
Grade 8 and below		2.458985	.396029	5.59	0.000	1.789997 3.377997
Some High School		2.328467	.2472903	7.96	0.000	1.88857 2.870828
Some College		.652351	.0675612	-4.12	0.000	.5318663 .8001292
College		.2732038	.0457529	-7.75	0.000	.1963772 .3800864
Advanced Degree		.1466958	.0581111	-4.85	0.000	.0671778 .3203389
region						

	Midwest		.4615236	.0820743	-4.35	0.000	.3250296	.6553372
	South		.2418332	.037658	-9.12	0.000	.1779019	.328739
	West		.5410815	.1034954	-3.21	0.002	.3710925	.7889385
	healthstatus							
	Very Good		1.146407	.1514986	1.03	0.302	.8834537	1.487626
	Good		1.600454	.1888955	3.98	0.000	1.268184	2.019779
	Fair		2.627319	.3440688	7.38	0.000	2.029453	3.401313
	Poor		8.334367	1.675319	10.55	0.000	5.607299	12.38773
	_cons		.0139927	.0034642	-17.24	0.000	.0085884	.0227975

Uninsured								
	agecat							
	18-24		1.012308	.0751319	0.16	0.869	.8745062	1.171825
	45-64		.8352937	.0600446	-2.50	0.013	.724916	.9624777
	sex							
	female		.7114564	.0371344	-6.52	0.000	.6418833	.7885705
	race							
	Black		1.290982	.1079286	3.05	0.003	1.094801	1.522316
	Hispanic		2.584553	.251933	9.74	0.000	2.132655	3.132205
	Asian		1.388923	.2369184	1.93	0.055	.9922482	1.944177
	Other		1.026743	.1731964	0.16	0.876	.736248	1.431857
	income							
	Poor		12.20298	1.337564	22.82	0.000	9.831327	15.14677
	Near Poor		6.679739	1.021245	12.42	0.000	4.941389	9.029629
	Low Income		4.116429	.4060274	14.35	0.000	3.388933	5.000094
	Middle Income		1.973032	.1945314	6.89	0.000	1.624469	2.396387
	notmsa		1.013337	.0948572	0.14	0.888	.8425598	1.218728
	education							
	Grade 8 and below		2.385928	.3866396	5.37	0.000	1.733409	3.28408
	Some High School		1.529818	.13387	4.86	0.000	1.287393	1.817893
	Some College		.7176392	.0604468	-3.94	0.000	.6078321	.8472833
	College		.3641578	.0364297	-10.10	0.000	.2989725	.4435556
	Advanced Degree		.2413696	.0407848	-8.41	0.000	.1729812	.3367954
	region							
	Midwest		1.029522	.1171758	0.26	0.798	.8225842	1.28852
	South		1.44036	.1495748	3.51	0.001	1.173688	1.767622
	West		1.167026	.1357711	1.33	0.186	.927821	1.467902
	healthstatus							
	Very Good		1.075455	.094229	0.83	0.407	.9048338	1.278249
	Good		.9787856	.0974768	-0.22	0.830	.8042914	1.191137
	Fair		1.118509	.1441952	0.87	0.386	.8674675	1.442201
	Poor		1.547575	.3166361	2.13	0.034	1.033857	2.316557
	_cons		.1109677	.0161457	-15.11	0.000	.0832937	.1478363

Public_Ins_Medicare								
	agecat							
	18-24		.3272802	.1272885	-2.87	0.005	.1520189	.7045989
	45-64		2.7896	.4287907	6.67	0.000	2.06028	3.777092
	sex							
	female		.7766956	.0915973	-2.14	0.033	.6155604	.980011
	race							

Black		1.0624	.1585739	0.41	0.686	.7915589	1.425913
Hispanic		.5384293	.112099	-2.97	0.003	.3571571	.8117048
Asian		.5821763	.4028655	-0.78	0.435	.1487736	2.278155
Other		1.519349	.666832	0.95	0.342	.6395171	3.60963
income							
Poor		31.862	8.883601	12.41	0.000	18.38806	55.20903
Near Poor		19.46689	6.816125	8.48	0.000	9.760842	38.82448
Low Income		9.414555	2.612306	8.08	0.000	5.447658	16.27008
Middle Income		3.352517	.8790622	4.61	0.000	1.999182	5.621985
notmsa							
		1.006533	.1704649	0.04	0.969	.7207981	1.405537
education							
Grade 8 and below		2.501026	.5636129	4.07	0.000	1.603831	3.900119
Some High School		1.920221	.3673139	3.41	0.001	1.316923	2.799898
Some College		1.061705	.2286997	0.28	0.781	.6943191	1.623485
College		.493785	.1643647	-2.12	0.035	.2561642	.9518256
Advanced Degree		.215114	.124479	-2.66	0.009	.0687358	.6732164
region							
Midwest		.5402748	.1240104	-2.68	0.008	.3436184	.8494799
South		.6407921	.1378503	-2.07	0.040	.4192906	.9793076
West		.5101346	.1481795	-2.32	0.021	.2877178	.9044879
healthstatus							
Very Good		1.68151	.5728354	1.53	0.129	.8590149	3.291531
Good		3.508591	1.075593	4.09	0.000	1.917074	6.421353
Fair		8.590233	2.618365	7.06	0.000	4.709901	15.66744
Poor		24.18487	8.456658	9.11	0.000	12.13777	48.18908
_cons		.0027171	.0012433	-12.91	0.000	.0011023	.0066974

```
.
. tabulate education
```

education	Freq.	Percent	Cum.
-----+-----			
Grade 8 and below	7,504	27.54	27.54
Some High School	3,732	13.70	41.24
HighSchool	7,544	27.69	68.93
Some College	3,629	13.32	82.25
College	3,245	11.91	94.16
Advanced Degree	1,590	5.84	100.00
-----+-----			
Total	27,244	100.00	

```
. margins, dydx(1.income 2.income) predict(outcome(3)) noestimcheck/*uninsured*/
```

```
Average marginal effects      Number of obs   =      26721
Model VCE      : Linearized
```

```
Expression      : Pr(insurance==Uninsured), predict(outcome(3))
dy/dx w.r.t.    : 1.income 2.income
```

		Delta-method				
		dy/dx	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
income						
Poor		.2638672	.0177408	14.87	0.000	.2290959 .2986385
Near Poor		.2173327	.0245804	8.84	0.000	.169156 .2655094

	agecat						
	18-24		-.0399506	.0417979	-0.96	0.340	-.1223594 .0424582
	45-64		.3219381	.0328259	9.81	0.000	.2572184 .3866578
	sex						
	female		.5060721	.0291953	17.33	0.000	.4485106 .5636336
	race						
	Black		-.4122867	.0428795	-9.62	0.000	-.496828 -.3277453
	Hispanic		-.4398107	.0423192	-10.39	0.000	-.5232472 -.3563741
	Asian		-.4599805	.0655998	-7.01	0.000	-.5893174 -.3306436
	Other		-.185569	.1248124	-1.49	0.139	-.4316496 .0605115
	income						
	Poor		-.2847111	.0525734	-5.42	0.000	-.3883651 -.1810571
	Near Poor		-.297532	.0697982	-4.26	0.000	-.4351465 -.1599176
	Low Income		-.1792278	.0558786	-3.21	0.002	-.2893982 -.0690574
	Middle Income		-.2011422	.0400596	-5.02	0.000	-.2801238 -.1221605
	notmsa		-.0567939	.038795	-1.46	0.145	-.1332823 .0196944
	education						
	Grade 8 and below		-.0293005	.0641882	-0.46	0.649	-.1558542 .0972532
	Some High School		-.0381313	.0491379	-0.78	0.439	-.1350118 .0587491
	Some College		.2053192	.0403966	5.08	0.000	.1256732 .2849652
	College		.3229223	.047472	6.80	0.000	.2293263 .4165183
	Advanced Degree		.3360283	.0682553	4.92	0.000	.201456 .4706007
	region						
	Midwest		.0307644	.0585583	0.53	0.600	-.0846894 .1462182
	South		-.0436228	.0507469	-0.86	0.391	-.1436757 .05643
	West		-.0942803	.0541906	-1.74	0.083	-.2011226 .012562
	healthstatus						
	Very Good		.201293	.0387591	5.19	0.000	.1248755 .2777106
	Good		.2759373	.0454517	6.07	0.000	.1863245 .3655501
	Fair		.8354909	.0585048	14.28	0.000	.7201426 .9508392
	Poor		1.386249	.1307244	10.60	0.000	1.128512 1.643986
	insurance						
	Public Ins not Medicare		.1483934	.0581607	2.55	0.011	.0337236 .2630633
	Uninsured		-.7853768	.0409731	-19.17	0.000	-.8661594 -.7045941
	Public Ins Medicare		.2045003	.1407427	1.45	0.148	-.0729885 .4819891
	_cons		.7679485	.0644321	11.92	0.000	.6409139 .8949831

```
. outreg2 using "hw6twopartmodel.xls", replace
hw6twopartmodel.xls
dir : seeout
```

```
. margins, dydx(1.income 2.income 3.income 4.income 5.income) noestimcheck
```

```
Average marginal effects          Number of obs   =       26721
Model VCE      : Linearized
```

```
Expression      : Pr(healthexpn), predict()
dy/dx w.r.t.    : 1.income 2.income 3.income 4.income
```

		Delta-method			
		dy/dx	Std. Err.	z	P> z
					[95% Conf. Interval]

income							
Poor		-.0637865	.0121337	-5.26	0.000	-.0875682	-.0400049
Near Poor		-.0669817	.0167173	-4.01	0.000	-.099747	-.0342164
Low Income		-.0385407	.0122193	-3.15	0.002	-.0624901	-.0145913
Middle Income		-.0436311	.0086933	-5.02	0.000	-.0606697	-.0265925

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

.
.
.

. predict probithealthexpnhat
(option pr assumed; Pr(healthexpn))
(6345 missing values generated)

. summarize probithealthexpnhat, detail

Pr(healthexpn)					
Percentiles		Smallest			
1%	.2413384	.1679557			
5%	.3767758	.1679557			
10%	.4997447	.1769287	Obs	26721	
25%	.6762185	.1769287	Sum of Wgt.	26721	
50%	.8179805		Mean	.7707458	
		Largest	Std. Dev.	.1821505	
75%	.9115369	.9993678			
90%	.9598115	.9994083	Variance	.0331788	
95%	.9759533	.9994459	Skewness	-1.105731	
99%	.9931905	.9995749	Kurtosis	3.635765	

.
. gen age18to64health = 0 if age08x !=-1
(251 missing values generated)

. replace age18to64health = 1 if age08x >= 18 & age08x <=64 & healthexp > 0
(15035 real changes made)

.
. svy, subpop(age18to64health): reg loghealthexp ib3.agecat ib1.sex ib1.race ib5.income notmsa
ib3.education ib1.region ib1.health
> status ib1.insurance
(running regress on estimation sample)

Survey: Linear regression

Number of strata	=	165	Number of obs	=	31391
Number of PSUs	=	370	Population size	=	285006026
			Subpop. no. of obs	=	13000
			Subpop. size	=	138614555
			Design df	=	205
			F(27, 179)	=	117.96
			Prob > F	=	0.0000
			R-squared	=	0.1995

		Linearized				
loghealthexp		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
agecat						
18-24		-.0968416	.0533238	-1.82	0.071	-.201975 .0082918

45-64		.5160689	.0322822	15.99	0.000	.4524212	.5797166
sex							
female		.4284671	.0324599	13.20	0.000	.364469	.4924653
race							
Black		-.4071942	.0475736	-8.56	0.000	-.5009905	-.313398
Hispanic		-.4345545	.042628	-10.19	0.000	-.5186001	-.3505089
Asian		-.6676905	.0684702	-9.75	0.000	-.8026867	-.5326943
Other		.0075954	.1025165	0.07	0.941	-.1945266	.2097173
income							
Poor		-.1502753	.0599618	-2.51	0.013	-.2684962	-.0320545
Near Poor		-.2352077	.08588	-2.74	0.007	-.4045289	-.0658864
Low Income		-.1893176	.0603854	-3.14	0.002	-.3083737	-.0702614
Middle Income		-.0933554	.0388537	-2.40	0.017	-.1699594	-.0167513
notmsa		-.126314	.0532493	-2.37	0.019	-.2313004	-.0213276
education							
Grade 8 and below		-.1765067	.0776802	-2.27	0.024	-.3296612	-.0233522
Some High School		-.1327232	.0579384	-2.29	0.023	-.2469547	-.0184917
Some College		.0924306	.040515	2.28	0.024	.0125512	.1723101
College		.1799965	.0465127	3.87	0.000	.0882918	.2717012
Advanced Degree		.2815982	.0585458	4.81	0.000	.1661692	.3970272
region							
Midwest		.0490861	.0475085	1.03	0.303	-.0445818	.1427541
South		.0121505	.0468241	0.26	0.796	-.0801681	.1044691
West		-.0073031	.0476285	-0.15	0.878	-.1012077	.0866016
healthstatus							
Very Good		.2757654	.0406343	6.79	0.000	.1956506	.3558802
Good		.5885754	.0405297	14.52	0.000	.5086668	.668484
Fair		1.122856	.0584263	19.22	0.000	1.007663	1.23805
Poor		1.850241	.0847337	21.84	0.000	1.683179	2.017302
insurance							
Public Ins not Medicare		.1275193	.0748825	1.70	0.090	-.0201194	.275158
Uninsured		-.7144227	.0494021	-14.46	0.000	-.811824	-.6170213
Public Ins Medicare		.7523009	.0985848	7.63	0.000	.5579307	.946671
_cons		6.427621	.0652088	98.57	0.000	6.299055	6.556187

```

. outreg2 using "hw6twopartmodel.xls", replace append
replaced when both replace and append chosen
hw6twopartmodel.xls
dir : seeout

```

```

. margins, dydx(1.income 2.income 3.income 4.income 5.income education) noestimcheck

```

```

Average marginal effects      Number of obs   =      26618
Model VCE      : Linearized

```

```

Expression      : Linear prediction, predict()
dy/dx w.r.t.    : 1.income 2.income 3.income 4.income 1.education 2.education 4.education
5.education 6.education

```

		Delta-method				
		dy/dx	Std. Err.	t	P> t	[95% Conf. Interval]

income							
Poor		-.1502753	.0599618	-2.51	0.013	-.2684962	-.0320545
Near Poor		-.2352077	.08588	-2.74	0.007	-.4045289	-.0658864
Low Income		-.1893176	.0603854	-3.14	0.002	-.3083737	-.0702614
Middle Income		-.0933554	.0388537	-2.40	0.017	-.1699594	-.0167513
education							
Grade 8 and below		-.1765067	.0776802	-2.27	0.024	-.3296612	-.0233522
Some High School		-.1327232	.0579384	-2.29	0.023	-.2469547	-.0184917
Some College		.0924306	.040515	2.28	0.024	.0125512	.1723101
College		.1799965	.0465127	3.87	0.000	.0882918	.2717012
Advanced Degree		.2815982	.0585458	4.81	0.000	.1661692	.3970272

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

```
.
. predict healthexphat
(option xb assumed; fitted values)
(6345 missing values generated)
```

```
. summarize healthexphat, detail
```

Fitted values				

	Percentiles	Smallest		
1%	5.067243	4.638776		
5%	5.531896	4.679684		
10%	5.782122	4.691834	Obs	26721
25%	6.214987	4.711423	Sum of Wgt.	26721
50%	6.715537		Mean	6.764035
		Largest	Std. Dev.	.7934478
75%	7.278801	9.966471		
90%	7.792741	9.968574	Variance	.6295595
95%	8.114679	9.974698	Skewness	.3315561
99%	8.827353	10.05404	Kurtosis	3.18395

```
. predict resid, residuals
(6526 missing values generated)
```

```
. gen exphealthexphat = exp(healthexphat)
(6345 missing values generated)
```

```
. gen residsmear = exp(resid)
(6526 missing values generated)
```

```
. egen avgresidsmear = mean(residsmear)
```

```
. gen smear = avgresidsmear*exphealthexphat
(6345 missing values generated)
```

```
. summarize smear
```

Variable		Obs	Mean	Std. Dev.	Min	Max
-----+-----						
smear		26721	2756.259	3067.961	233.0482	52391.85

```
.
.
.
. /*Part B*/
.
. /*Park Test*/
```

```
. svy, subpop(age18to64): glm healthexp ib3.agecat ib1.sex ib1.race ib5.income notmsa
ib3.education ib1.region ib1.healthstatus ib
> 1.insurance, family(gamma) link(log)
(running glm on estimation sample)
```

Survey: Generalized linear models

Number of strata	=	165	Number of obs	=	31113
Number of PSUs	=	370	Population size	=	282464364
			Subpop. no. of obs	=	16917
			Subpop. size	=	170277348
			Design df	=	205

		Linearized					
healthexp		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
agecat							
	18-24	-.1525881	.0708288	-2.15	0.032	-.2922343	-.0129418
	45-64	.4421097	.0433598	10.20	0.000	.3566214	.527598
sex							
	female	.4685999	.0384777	12.18	0.000	.392737	.5444627
race							
	Black	-.4413024	.0592257	-7.45	0.000	-.5580719	-.3245329
	Hispanic	-.5400887	.0616302	-8.76	0.000	-.661599	-.4185783
	Asian	-.7675889	.0722038	-10.63	0.000	-.9099462	-.6252317
	Other	.0192808	.1349877	0.14	0.887	-.2468613	.285423
income							
	Poor	-.1192354	.088956	-1.34	0.182	-.2946214	.0561505
	Near Poor	-.1538654	.093684	-1.64	0.102	-.3385731	.0308423
	Low Income	-.2147216	.0717328	-2.99	0.003	-.3561503	-.073293
	Middle Income	-.1369234	.0433445	-3.16	0.002	-.2223815	-.0514653
	notmsa	-.1861548	.0524585	-3.55	0.000	-.2895822	-.0827274
education							
	Grade 8 and below	-.0523859	.1500581	-0.35	0.727	-.3482409	.2434692
	Some High School	-.0763148	.0736007	-1.04	0.301	-.2214262	.0687966
	Some College	.1285256	.0481397	2.67	0.008	.0336131	.2234381
	College	.2032447	.0560938	3.62	0.000	.0926499	.3138394
	Advanced Degree	.2645275	.0751969	3.52	0.001	.1162689	.412786
region							
	Midwest	.0598253	.0560905	1.07	0.287	-.050763	.1704136
	South	.0173792	.0572827	0.30	0.762	-.0955596	.130318
	West	-.0469751	.0534701	-0.88	0.381	-.152397	.0584468
healthstatus							
	Very Good	.3019975	.0573575	5.27	0.000	.1889112	.4150838
	Good	.6460287	.0520156	12.42	0.000	.5434745	.7485829
	Fair	1.192653	.0652875	18.27	0.000	1.063932	1.321374
	Poor	1.740819	.0787071	22.12	0.000	1.58564	1.895998
insurance							
	Public Ins not Medicare	.3835083	.0833456	4.60	0.000	.2191839	.5478327
	Uninsured	-.9190729	.0725297	-12.67	0.000	-1.062073	-.7760731
	Public Ins Medicare	.7460566	.0923205	8.08	0.000	.5640372	.928076
	_cons	7.157574	.0815598	87.76	0.000	6.996771	7.318378

```
. predict healthhat2, xb
(6345 missing values generated)

. gen lnhealthhat2 = ln(healthhat2)
(6345 missing values generated)

. gen r2 = ((healthexp-healthhat2))^2
(6526 missing values generated)

. gen lnr2 = ln(r2)
(6526 missing values generated)

.
.
. svy, subpop(age18to64): glm lnr2 lnhealthhat2, family(gamma) link(log) nolog
(running glm on estimation sample)
```

Survey: Generalized linear models

Number of strata	=	165	Number of obs	=	31113
Number of PSUs	=	370	Population size	=	282464364
			Subpop. no. of obs	=	16917
			Subpop. size	=	170277348
			Design df	=	205

	lnr2	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnhealthhat2		2.230204	.0444583	50.16	0.000	2.142549	2.317858
_cons		-2.048806	.0920827	-22.25	0.000	-2.230357	-1.867256

```
. test lnhealthhat2 == 0
```

Adjusted Wald test

```
( 1)  [lnr2]lnhealthhat2 = 0

      F( 1, 205) = 2516.42
      Prob > F = 0.0000
```

```
. test lnhealthhat2 == 1
```

Adjusted Wald test

```
( 1)  [lnr2]lnhealthhat2 = 1

      F( 1, 205) = 765.68
      Prob > F = 0.0000
```

```
. test lnhealthhat2 == 2
```

Adjusted Wald test

```
( 1)  [lnr2]lnhealthhat2 = 2

      F( 1, 205) = 26.81
      Prob > F = 0.0000
```

```
. test lnhealthhat2 == 3
```

Adjusted Wald test

(1) [lnr2]lnhealthhat2 = 3

F(1, 205) = 299.81
Prob > F = 0.0000

```
.
.
. /*Coefficient=2.2 Gamma*/
.
. /*Part C*/
.
. svy, subpop(age18to64): glm healthexp ib3.agecat ib1.sex ib1.race ib5.income notmsa
ib3.education ib1.region ib1.healthstatus ib
> 1.insurance, family(gamma) link(log)
(running glm on estimation sample)
```

Survey: Generalized linear models

Number of strata	=	165	Number of obs	=	31113
Number of PSUs	=	370	Population size	=	282464364
			Subpop. no. of obs	=	16917
			Subpop. size	=	170277348
			Design df	=	205

		Linearized					
healthexp		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
agecat							
18-24		-.1525881	.0708288	-2.15	0.032	-.2922343	-.0129418
45-64		.4421097	.0433598	10.20	0.000	.3566214	.527598
sex							
female		.4685999	.0384777	12.18	0.000	.392737	.5444627
race							
Black		-.4413024	.0592257	-7.45	0.000	-.5580719	-.3245329
Hispanic		-.5400887	.0616302	-8.76	0.000	-.661599	-.4185783
Asian		-.7675889	.0722038	-10.63	0.000	-.9099462	-.6252317
Other		.0192808	.1349877	0.14	0.887	-.2468613	.285423
income							
Poor		-.1192354	.088956	-1.34	0.182	-.2946214	.0561505
Near Poor		-.1538654	.093684	-1.64	0.102	-.3385731	.0308423
Low Income		-.2147216	.0717328	-2.99	0.003	-.3561503	-.073293
Middle Income		-.1369234	.0433445	-3.16	0.002	-.2223815	-.0514653
notmsa							
notmsa		-.1861548	.0524585	-3.55	0.000	-.2895822	-.0827274
education							
Grade 8 and below		-.0523859	.1500581	-0.35	0.727	-.3482409	.2434692
Some High School		-.0763148	.0736007	-1.04	0.301	-.2214262	.0687966
Some College		.1285256	.0481397	2.67	0.008	.0336131	.2234381
College		.2032447	.0560938	3.62	0.000	.0926499	.3138394
Advanced Degree		.2645275	.0751969	3.52	0.001	.1162689	.412786
region							
Midwest		.0598253	.0560905	1.07	0.287	-.050763	.1704136
South		.0173792	.0572827	0.30	0.762	-.0955596	.130318
West		-.0469751	.0534701	-0.88	0.381	-.152397	.0584468

healthstatus							
Very Good		.3019975	.0573575	5.27	0.000	.1889112	.4150838
Good		.6460287	.0520156	12.42	0.000	.5434745	.7485829
Fair		1.192653	.0652875	18.27	0.000	1.063932	1.321374
Poor		1.740819	.0787071	22.12	0.000	1.58564	1.895998
insurance							
Public Ins not Medicare		.3835083	.0833456	4.60	0.000	.2191839	.5478327
Uninsured		-.9190729	.0725297	-12.67	0.000	-1.062073	-.7760731
Public Ins Medicare		.7460566	.0923205	8.08	0.000	.5640372	.928076
_cons		7.157574	.0815598	87.76	0.000	6.996771	7.318378

```
. outreg2 using "hw6glm.xls"
hw6glm.xls
dir : seeout
```

```
. margins, dydx(1.income 2.income 3.income 4.income 5.income) noestimcheck
```

```
Average marginal effects      Number of obs   =      26618
Model VCE      : Linearized
```

```
Expression      : Predicted mean healthexp, predict()
dy/dx w.r.t.    : 1.income 2.income 3.income 4.income
```

		Delta-method				[95% Conf. Interval]	
		dy/dx	Std. Err.	z	P> z		
income							
Poor		-363.8902	263.5997	-1.38	0.167	-880.5362	152.7558
Near Poor		-461.6971	268.4587	-1.72	0.085	-987.8665	64.47223
Low Income		-625.5801	202.7473	-3.09	0.002	-1022.958	-228.2027
Middle Income		-414.2702	132.3813	-3.13	0.002	-673.7328	-154.8077

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

```
.
.
. /*Part E*/
.
. quietly svy, subpop(age18to64): glm healthexp ib3.agecat ib1.sex ib1.race ib5.income notmsa
ib3.education ib1.region ib1.healths
> tatus ib1.insurance, family(gamma) link(log)

. predict glmexppredict
(option mu assumed; predicted mean healthexp)
(6345 missing values generated)

. summarize glmexppredict, detail
```

Predicted mean healthexp			
Percentiles	Smallest		
1%	251.2586	155.1926	
5%	422.3792	157.2931	
10%	607.9389	161.7436	Obs 26721
25%	1076.563	170.8991	Sum of Wgt. 26721
50%	1828.039		Mean 2625.364
		Largest	Std. Dev. 2796.034
75%	3158.811	39124.57	

90%	5215.802	41121.04	Variance	7817808
95%	7386.246	41121.04	Skewness	4.001164
99%	14104.01	42803.04	Kurtosis	29.73515

```
.
. gen predictolsmodel = smear*healthexphat*probithealthexpnhat
(6345 missing values generated)
```

```
. summarize predictolsmodel, detail
```

```

-----+-----
               predictolsmodel
-----+-----
Percentiles      Smallest
1%           440.6569    200.7225
5%           1134.039    200.7225
10%          2200.526    221.8453      Obs           26721
25%          4875.182    225.2032      Sum of Wgt.      26721

50%          10272.64
               Largest      Mean           18453.2
75%          21501.58      477914.6      Std. Dev.      27925.07
90%          40542.86      479207.3      Variance       7.80e+08
95%          58902.71      482385.7      Skewness       5.665885
99%          134717.8      526405.1      Kurtosis       56.15866
```

```
.
end of do-file
```