

```
-----
name: <unnamed>
log: D:\Jason\workshop\regression analysis 2022\regression2.log
log type: text
opened on: 3 Oct 2022, 10:52:37
```

```
. webuse highschool, clear
```

```
. cd "D:\Jason\workshop\regression analysis 2022"
D:\Jason\workshop\regression analysis 2022
```

```
.
. *****
. * Continuous dependent variable
. *****
.
. *****
. * Questions 1: Determine if X2 is an important predictor when X1 is already in the model
. *****
```

```
.
. *****
. * 1.1 Regression without complex survey data
. *****
```

```
.
. *****
. * 1.1.0 Regression without using the weight variable
. *****
```

```
.
. reg weight height i.race
```

Source	SS	df	MS	Number of obs	=	4,071
Model	1166634.57	3	388878.19	F(3, 4067)	=	432.31
Residual	3658378.15	4,067	899.527452	Prob > F	=	0.0000
Total	4825012.72	4,070	1185.50681	R-squared	=	0.2418
				Adj R-squared	=	0.2412
				Root MSE	=	29.992

weight	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
height	.6609619	.0188005	35.16	0.000	.6241025	.6978212
race						
Black	7.697073	1.500755	5.13	0.000	4.754771	10.63938
Other	-12.28	3.654374	-3.36	0.001	-19.44457	-5.115422
_cons	-125.0944	8.110502	-15.42	0.000	-140.9954	-109.1934

```
.
. *****
. * 1.1.1. Regression using a personal weight variable for the whole sample
. *****
```

```
.
. reg weight height i.race [pw=sampwgt]
(sum of wgt is 8,000,000)
```

Linear regression	Number of obs	=	4,071
	F(3, 4067)	=	376.04
	Prob > F	=	0.0000
	R-squared	=	0.2839
	Root MSE	=	29.051



```

-----+-----
height | .4122218 .0613759 6.72 0.000 .2733797 .5510638
  race |
Black  | 13.97939 3.113373 4.49 0.002 6.936449 21.02233
Other  | -6.629714 2.649844 -2.50 0.034 -12.62408 -.6353504
  _cons | -26.49653 25.33738 -1.05 0.323 -83.81367 30.82062
-----+-----
    
```

```

.
.
. *****
. * 1.2. Regression with complex survey data
. *****
. svydes
.
    
```

Survey: Describing stage 1 sampling units

```

pweight: sampwgt
VCE: linearized
Single unit: missing
Strata 1: state
SU 1: county
FPC 1: ncounties
Strata 2: <one>
SU 2: school
FPC 2: nschools
    
```

```

#Obs per Unit
-----+-----
Stratum  #Units  #Obs  min  mean  max
-----+-----
1         2        92   34   46.0  58
2         2       112   51   56.0  61
3         2        43   18   21.5  25
4         2        37   14   18.5  23
5         2        96   38   48.0  58
6         2        76   27   38.0  49
7         2        97   47   48.5  50
8         2       103   51   51.5  52
9         2        69   33   34.5  36
10        2        50   17   25.0  33
11        2        72   30   36.0  42
12        2        76   34   38.0  42
13        2       135   54   67.5  81
14        2       103   43   51.5  60
15        2        66   28   33.0  38
16        2        72   31   36.0  41
17        2        60   26   30.0  34
18        2        58   25   29.0  33
19        2        72   35   36.0  37
20        2        83   38   41.5  45
21        2        80   32   40.0  48
22        2        76   37   38.0  39
23        2        81   38   40.5  43
24        2        72   36   36.0  36
25        2        74   34   37.0  40
26        2        89   33   44.5  56
27        2       107   47   53.5  60
28        2       126   58   63.0  68
29        2        90   42   45.0  48
30        2        87   42   43.5  45
    
```

31	2	84	38	42.0	46
32	2	95	40	47.5	55
33	2	97	44	48.5	53
34	2	82	37	41.0	45
35	2	98	36	49.0	62
36	2	95	42	47.5	53
37	2	53	23	26.5	30
38	2	75	37	37.5	38
39	2	61	28	30.5	33
40	2	54	25	27.0	29
41	2	73	28	36.5	45
42	2	79	36	39.5	43
43	2	108	48	54.0	60
44	2	76	37	38.0	39
45	2	107	45	53.5	62
46	2	115	56	57.5	59
47	2	67	28	33.5	39
48	2	56	23	28.0	33
49	2	78	39	39.0	39
50	2	64	31	32.0	33

-----  
 50            100            4,071            14            40.7            81  
 -----

```
. svyset, clear
. quietly svyset county [pw = sampwgt], fpc(ncounties) strata(state)|| school, fpc(nschools)
```

```
*****
* 1.2.1. Regression using complex survey data for the whole sample
*****
```

```
. svy: reg weight height i.race
(running regress on estimation sample)
```

Survey: Linear regression

```
Number of strata = 50                      Number of obs = 4,071
Number of PSUs = 100                      Population size = 8,000,000
Design df = 50
F( 3, 48) = 207.69
Prob > F = 0.0000
R-squared = 0.2839
```

```
-----
```

	Coef.	Linearized Std. Err.	t	P> t	[95% Conf. Interval]	
weight						
height	.7132085	.029109	24.50	0.000	.6547413	.7716758
race						
Black	6.90636	2.477742	2.79	0.007	1.929669	11.88305
Other	-8.328094	2.557726	-3.26	0.002	-13.46544	-3.190751
_cons	-148.7428	12.3825	-12.01	0.000	-173.6138	-123.8719

```
-----
```

```
*****
* 1,2,2. Regression using complex survey data and adjusted for
* dependent observations from the same schools.
```











Black	-14.95026	62.69483	-0.24	0.817	-156.7758	126.8753
Other	-163.9238	47.83865	-3.43	0.008	-272.1424	-55.7053
race#c.height						
Black	.0288018	.1417587	0.20	0.844	-.2918788	.3494823
Other	.3403916	.1064657	3.20	0.011	.0995494	.5812338
sex						
female	136.4661	42.54978	3.21	0.011	40.21179	232.7204
sex#c.height						
female	-.3405318	.1005027	-3.39	0.008	-.5678846	-.1131789
race#sex						
Black#female	8.525301	104.1438	0.08	0.937	-227.0643	244.1149
Other#female	13.26205	178.4405	0.07	0.942	-390.3983	416.9224
race#sex#c.height						
Black#female	.0202906	.2453719	0.08	0.936	-.5347792	.5753605
Other#female	.0183652	.437561	0.04	0.967	-.9714665	1.008197
_cons	-157.8568	23.37035	-6.75	0.000	-210.7243	-104.9894

. reg, coeflegend

```

Linear regression      Number of obs   =      4,071
                       F(8, 9)                =            .
                       Prob > F                =            .
                       R-squared                =      0.3057
                       Root MSE              =      28.633
    
```

(Std. Err. adjusted for 10 clusters in school)

weight	Coef.	Legend
height	.740487	_b[height]
race		
Black	-14.95026	_b[2.race]
Other	-163.9238	_b[3.race]
race#c.height		
Black	.0288018	_b[2.race#c.height]
Other	.3403916	_b[3.race#c.height]
sex		
female	136.4661	_b[2.sex]
sex#c.height		
female	-.3405318	_b[2.sex#c.height]
race#sex		
Black#female	8.525301	_b[2.race#2.sex]
Other#female	13.26205	_b[3.race#2.sex]
race#sex#c.height		
Black#female	.0202906	_b[2.race#2.sex#c.height]
Other#female	.0183652	_b[3.race#2.sex#c.height]
_cons	-157.8568	_b[_cons]

```

.
.      * test the coefficients of two-way interactions
.      test _b[2.race#c.height] = _b[3.race#c.height]

( 1) 2.race#c.height - 3.race#c.height = 0

      F( 1,      9) =      5.14
      Prob > F =      0.0496

.      test _b[2.race#c.height] = -_b[3.race#c.height]

( 1) 2.race#c.height + 3.race#c.height = 0

      F( 1,      9) =      3.10
      Prob > F =      0.1122

.
.      * test the coefficients of three-way interactions
.      test _b[2.race#2.sex#c.height] = _b[3.race#2.sex#c.height]

( 1) 2.race#2.sex#c.height - 3.race#2.sex#c.height = 0

      F( 1,      9) =      0.00
      Prob > F =      0.9973

.      test _b[2.race#2.sex#c.height] = -_b[3.race#2.sex#c.height]

( 1) 2.race#2.sex#c.height + 3.race#2.sex#c.height = 0

      F( 1,      9) =      0.01
      Prob > F =      0.9329

.
.
.      *****
.      * 4.2 A two-way interaction Modify the model 2.2.
.      *****
.
.      svy: reg weight c.height##i.race##i.sex
(running regress on estimation sample)
    
```

Survey: Linear regression

```

Number of strata =      50          Number of obs =      4,071
Number of PSUs  =     100          Population size = 8,000,000
                                   Design df      =      50
                                   F( 11,      40) =      65.77
                                   Prob > F        =      0.0000
                                   R-squared        =      0.3057
    
```

	weight	Coef.	Linearized Std. Err.	t	P> t	[95% Conf. Interval]	
	height	.740487	.0549802	13.47	0.000	.630056	.8509181
	race						
	Black	-14.95026	77.7808	-0.19	0.848	-171.1776	141.2771
	Other	-163.9238	72.88719	-2.25	0.029	-310.3221	-17.52561
	race#c.height						
	Black	.0288018	.1734996	0.17	0.869	-.3196824	.3772859

Other	.3403916	.1630919	2.09	0.042	.0128119	.6679713
sex						
female	136.4661	33.61345	4.06	0.000	68.9515	203.9807
sex#c.height						
female	-.3405318	.0793713	-4.29	0.000	-.4999537	-.1811098
race#sex						
Black#female	8.525301	102.9227	0.08	0.934	-198.201	215.2516
Other#female	13.26205	164.6361	0.08	0.936	-317.4193	343.9434
race#sex#c.height						
Black#female	.0202906	.2392789	0.08	0.933	-.4603152	.5008964
Other#female	.0183652	.3981457	0.05	0.963	-.7813339	.8180642
_cons	-157.8568	24.38347	-6.47	0.000	-206.8325	-108.8812

. reg, coeflegend

Survey: Linear regression

Number of strata	=	50	Number of obs	=	4,071
Number of PSUs	=	100	Population size	=	8,000,000
			Design df	=	50
			F( 11, 40)	=	65.77
			Prob > F	=	0.0000
			R-squared	=	0.3057

weight	Coef.	Legend
height	.740487	_b[height]
race		
Black	-14.95026	_b[2.race]
Other	-163.9238	_b[3.race]
race#c.height		
Black	.0288018	_b[2.race#c.height]
Other	.3403916	_b[3.race#c.height]
sex		
female	136.4661	_b[2.sex]
sex#c.height		
female	-.3405318	_b[2.sex#c.height]
race#sex		
Black#female	8.525301	_b[2.race#2.sex]
Other#female	13.26205	_b[3.race#2.sex]
race#sex#c.height		
Black#female	.0202906	_b[2.race#2.sex#c.height]
Other#female	.0183652	_b[3.race#2.sex#c.height]
_cons	-157.8568	_b[_cons]

. \* test the coefficients of two-way interactions  
 . test \_b[2.race#c.height] = \_b[3.race#c.height]

Adjusted Wald test

( 1) 2.race#c.height - 3.race#c.height = 0

F( 1, 50) = 1.89  
 Prob > F = 0.1749

. test \_b[2.race#c.height] = -\_b[3.race#c.height]

Adjusted Wald test

( 1) 2.race#c.height + 3.race#c.height = 0

F( 1, 50) = 2.19  
 Prob > F = 0.1449

. \* test the coefficients of three-way interactions  
 . test \_b[2.race#2.sex#c.height] = \_b[3.race#2.sex#c.height]

Adjusted Wald test

( 1) 2.race#2.sex#c.height - 3.race#2.sex#c.height = 0

F( 1, 50) = 0.00  
 Prob > F = 0.9966

. test \_b[2.race#2.sex#c.height] = -\_b[3.race#2.sex#c.height]

Adjusted Wald test

( 1) 2.race#2.sex#c.height + 3.race#2.sex#c.height = 0

F( 1, 50) = 0.01  
 Prob > F = 0.9358

. \*\*\*\*\*  
 . \* Questions 5: testing the total effect of a variable X with the margins command and complex survey data  
 . \*\*\*\*\*

. \* test the total effect of sex  
 . svy: reg weight c.height##i.race##i.sex  
 (running regress on estimation sample)

Survey: Linear regression

Number of strata	=	50	Number of obs	=	4,071
Number of PSUs	=	100	Population size	=	8,000,000
			Design df	=	50
			F( 11, 40)	=	65.77
			Prob > F	=	0.0000
			R-squared	=	0.3057

	weight	Coef.	Linearized Std. Err.	t	P> t	[95% Conf. Interval]
	height	.740487	.0549802	13.47	0.000	.630056 .8509181
	race					

Black	-14.95026	77.7808	-0.19	0.848	-171.1776	141.2771
Other	-163.9238	72.88719	-2.25	0.029	-310.3221	-17.52561
race#c.height						
Black	.0288018	.1734996	0.17	0.869	-.3196824	.3772859
Other	.3403916	.1630919	2.09	0.042	.0128119	.6679713
sex						
female	136.4661	33.61345	4.06	0.000	68.9515	203.9807
sex#c.height						
female	-.3405318	.0793713	-4.29	0.000	-.4999537	-.1811098
race#sex						
Black#female	8.525301	102.9227	0.08	0.934	-198.201	215.2516
Other#female	13.26205	164.6361	0.08	0.936	-317.4193	343.9434
race#sex#c.height						
Black#female	.0202906	.2392789	0.08	0.933	-.4603152	.5008964
Other#female	.0183652	.3981457	0.05	0.963	-.7813339	.8180642
_cons	-157.8568	24.38347	-6.47	0.000	-206.8325	-108.8812

. \* The total effect of sex for the whole sample  
 . margins, by(sex)  
 .

Predictive margins                                 Number of obs     =     4,071  
 Model VCE     : Linearized  
 Expression    : Linear prediction, predict()  
 over           : sex

	Margin	Delta-method Std. Err.	t	P> t	[95% Conf. Interval]	
sex						
male	175.4809	.9271741	189.26	0.000	173.6186	177.3432
female	146.204	.8358546	174.92	0.000	144.5251	147.8829

. marginsplot, by(sex)

Variables that uniquely identify margins: sex

. margins i.sex, at(height=(355(50)515)) atmeans

Adjusted predictions                                 Number of obs     =     4,071  
 Model VCE     : Linearized

Expression    : Linear prediction, predict()

1.\_at         : height             =            355  
               1.race             =       .8785304 (mean)  
               2.race             =       .097057 (mean)  
               3.race             =       .0244125 (mean)  
               1.sex              =       .4810027 (mean)  
               2.sex              =       .5189973 (mean)

2.\_at         : height             =            405

```

1.race      = .8785304 (mean)
2.race      = .097057  (mean)
3.race      = .0244125 (mean)
1.sex       = .4810027 (mean)
2.sex       = .5189973 (mean)

3._at      : height    = 455
1.race      = .8785304 (mean)
2.race      = .097057  (mean)
3.race      = .0244125 (mean)
1.sex       = .4810027 (mean)
2.sex       = .5189973 (mean)

4._at      : height    = 505
1.race      = .8785304 (mean)
2.race      = .097057  (mean)
3.race      = .0244125 (mean)
1.sex       = .4810027 (mean)
2.sex       = .5189973 (mean)
    
```

	Margin	Delta-method Std. Err.	t	P> t	[95% Conf. Interval]	
_at#sex						
1#male	103.5056	4.463539	23.19	0.000	94.54031	112.4709
1#female	121.0924	3.247045	37.29	0.000	114.5705	127.6143
2#male	141.0852	2.092828	67.41	0.000	136.8816	145.2888
2#female	141.7663	.8083906	175.37	0.000	140.1426	143.39
3#male	178.6648	1.030582	173.36	0.000	176.5948	180.7348
3#female	162.4402	2.673706	60.75	0.000	157.0699	167.8105
4#male	216.2444	3.157481	68.49	0.000	209.9024	222.5864
4#female	183.1141	5.479865	33.42	0.000	172.1075	194.1208

```

. marginsplot, yline(0)
Variables that uniquely identify margins: height sex

.
.
. margins i.sex, at(height=(355(50)515)) at(race ==1)

Predictive margins                                Number of obs   =    4,071
Model VCE      : Linearized

Expression    : Linear prediction, predict()

1._at        : height      =    355
2._at        : height      =    405
3._at        : height      =    455
4._at        : height      =    505
5._at        : race        =     1
    
```

	Margin	Delta-method Std. Err.	t	P> t	[95% Conf. Interval]	
_at#sex						

1#male		103.5056	4.463539	23.19	0.000	94.54031	112.4709
1#female		121.0924	3.247045	37.29	0.000	114.5705	127.6143
2#male		141.0852	2.092828	67.41	0.000	136.8816	145.2888
2#female		141.7663	.8083906	175.37	0.000	140.1426	143.39
3#male		178.6648	1.030582	173.36	0.000	176.5948	180.7348
3#female		162.4402	2.673706	60.75	0.000	157.0699	167.8105
4#male		216.2444	3.157481	68.49	0.000	209.9024	222.5864
4#female		183.1141	5.479865	33.42	0.000	172.1075	194.1208
5#male		162.507	1.048197	155.03	0.000	160.4016	164.6124
5#female		151.6457	1.688154	89.83	0.000	148.2549	155.0364

-----  
 . marginsplot, yline(0) saving("D:\Jason\workshop\regression analysis 2022\graph\_white.gph", replace)

Variables that uniquely identify margins: height sex  
 Multiple at() options specified:  
 \_atoption=1: height=(355(50)515)  
 \_atoption=2: race ==1  
 (file D:\Jason\workshop\regression analysis 2022\graph\_white.gph saved)

. margins i.sex, at(height=(355(50)515)) at(race ==2)

Predictive margins  
 Model VCE : Linearized  
 Number of obs = 4,071

Expression : Linear prediction, predict()

1. \_at : height = 355  
 2. \_at : height = 405  
 3. \_at : height = 455  
 4. \_at : height = 505  
 5. \_at : race = 2

		Delta-method				
		Margin	Std. Err.	t	P> t	[95% Conf. Interval]
_at#sex						
1#male		103.5056	4.463539	23.19	0.000	94.54031 112.4709
1#female		121.0924	3.247045	37.29	0.000	114.5705 127.6143
2#male		141.0852	2.092828	67.41	0.000	136.8816 145.2888
2#female		141.7663	.8083906	175.37	0.000	140.1426 143.39
3#male		178.6648	1.030582	173.36	0.000	176.5948 180.7348
3#female		162.4402	2.673706	60.75	0.000	157.0699 167.8105
4#male		216.2444	3.157481	68.49	0.000	209.9024 222.5864
4#female		183.1141	5.479865	33.42	0.000	172.1075 194.1208
5#male		160.0175	3.840411	41.67	0.000	152.3038 167.7312
5#female		166.46	4.283372	38.86	0.000	157.8566 175.0634

-----  
 . marginsplot, yline(0) saving("D:\Jason\workshop\regression analysis 2022\graph\_black.gph", replace)

Variables that uniquely identify margins: height sex  
 Multiple at() options specified:  
 \_atoption=1: height=(355(50)515)  
 \_atoption=2: race ==2  
 (file D:\Jason\workshop\regression analysis 2022\graph\_black.gph saved)

```
. margins i.sex, at(height=(355(50)515)) at(race ==3)
Predictive margins                                Number of obs   =       4,071
Model VCE    : Linearized
```

Expression : Linear prediction, predict()

```
1._at      : height      =       355
2._at      : height      =       405
3._at      : height      =       455
4._at      : height      =       505
5._at      : race        =         3
```

	Margin	Delta-method Std. Err.	t	P> t	[95% Conf. Interval]	
_at#sex						
1#male	103.5056	4.463539	23.19	0.000	94.54031	112.4709
1#female	121.0924	3.247045	37.29	0.000	114.5705	127.6143
2#male	141.0852	2.092828	67.41	0.000	136.8816	145.2888
2#female	141.7663	.8083906	175.37	0.000	140.1426	143.39
3#male	178.6648	1.030582	173.36	0.000	176.5948	180.7348
3#female	162.4402	2.673706	60.75	0.000	157.0699	167.8105
4#male	216.2444	3.157481	68.49	0.000	209.9024	222.5864
4#female	183.1141	5.479865	33.42	0.000	172.1075	194.1208
5#male	145.85	4.438492	32.86	0.000	136.935	154.7649
5#female	156.1962	10.10268	15.46	0.000	135.9043	176.488

```
. marginsplot, yline(0) saving("D:\Jason\workshop\regression analysis 2022\graph_other.gph", replace)
```

```
Variables that uniquely identify margins: height sex
Multiple at() options specified:
  _atoption=1: height=(355(50)515)
  _atoption=2: race ==3
(file D:\Jason\workshop\regression analysis 2022\graph_other.gph saved)
```

```
. margins i.sex, atmeans
```

```
Adjusted predictions                                Number of obs   =       4,071
Model VCE    : Linearized
```

Expression : Linear prediction, predict()

```
at      : height      = 432.6394 (mean)
1.race  :             = .8785304 (mean)
2.race  :             = .097057 (mean)
3.race  :             = .0244125 (mean)
1.sex   :             = .4810027 (mean)
2.sex   :             = .5189973 (mean)
```

	Margin	Delta-method Std. Err.	t	P> t	[95% Conf. Interval]	
--	--------	---------------------------	---	------	----------------------	--



```

sex |
male | 161.8587 1.017217 159.12 0.000 159.8156 163.9019
female | 153.1946 1.489519 102.85 0.000 150.2028 156.1864
    
```

. margins, dydx(i.sex) atmeans

Conditional marginal effects                      Number of obs     =     4,071  
 Model VCE     : Linearized

```

Expression     : Linear prediction, predict()
dy/dx w.r.t.   : 2.sex
at
   height       =   432.6394 (mean)
   1.race       =   .8785304 (mean)
   2.race       =   .097057 (mean)
   3.race       =   .0244125 (mean)
   1.sex        =   .4810027 (mean)
   2.sex        =   .5189973 (mean)
    
```

	dy/dx	Delta-method Std. Err.	t	P> t	[95% Conf. Interval]	
sex						
female	-8.664163	1.972424	-4.39	0.000	-12.62589	-4.702433

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

. margins, dydx(i.sex i.race) atmeans

Conditional marginal effects                      Number of obs     =     4,071  
 Model VCE     : Linearized

```

Expression     : Linear prediction, predict()
dy/dx w.r.t.   : 2.race 3.race 2.sex
at
   height       =   432.6394 (mean)
   1.race       =   .8785304 (mean)
   2.race       =   .097057 (mean)
   3.race       =   .0244125 (mean)
   1.sex        =   .4810027 (mean)
   2.sex        =   .5189973 (mean)
    
```

	dy/dx	Delta-method Std. Err.	t	P> t	[95% Conf. Interval]	
race						
Black	6.491153	3.199031	2.03	0.048	.0657108	12.9166
Other	-5.650381	6.262037	-0.90	0.371	-18.22805	6.92729
sex						
female	-8.664163	1.972424	-4.39	0.000	-12.62589	-4.702433

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

```

.
.
. log close
   name: <unnamed>
   log: D:\Jason\workshop\regression analysis 2022\regression2.log
   log type: text
closed on: 3 Oct 2022, 10:52:43
    
```