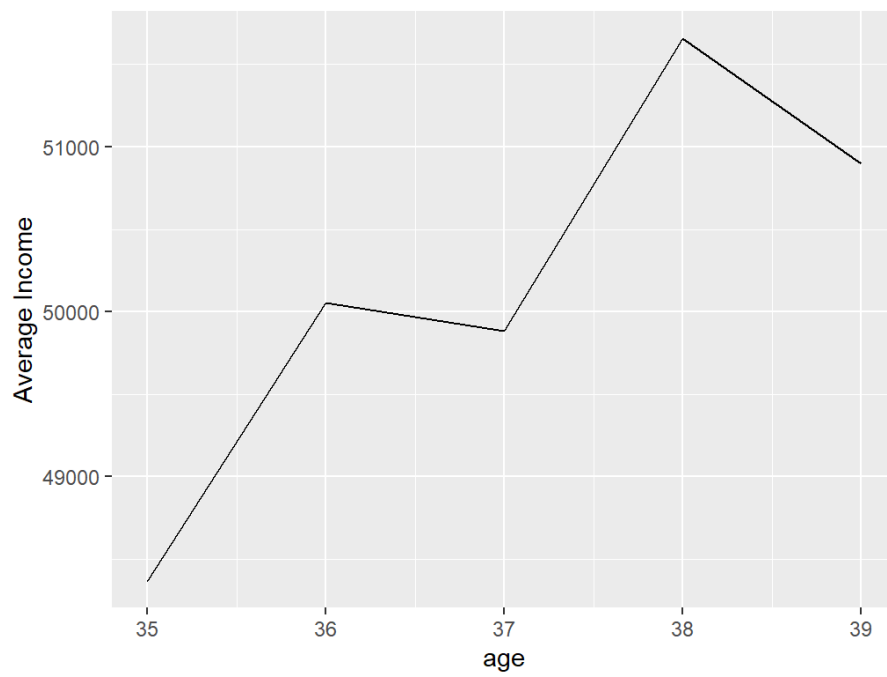
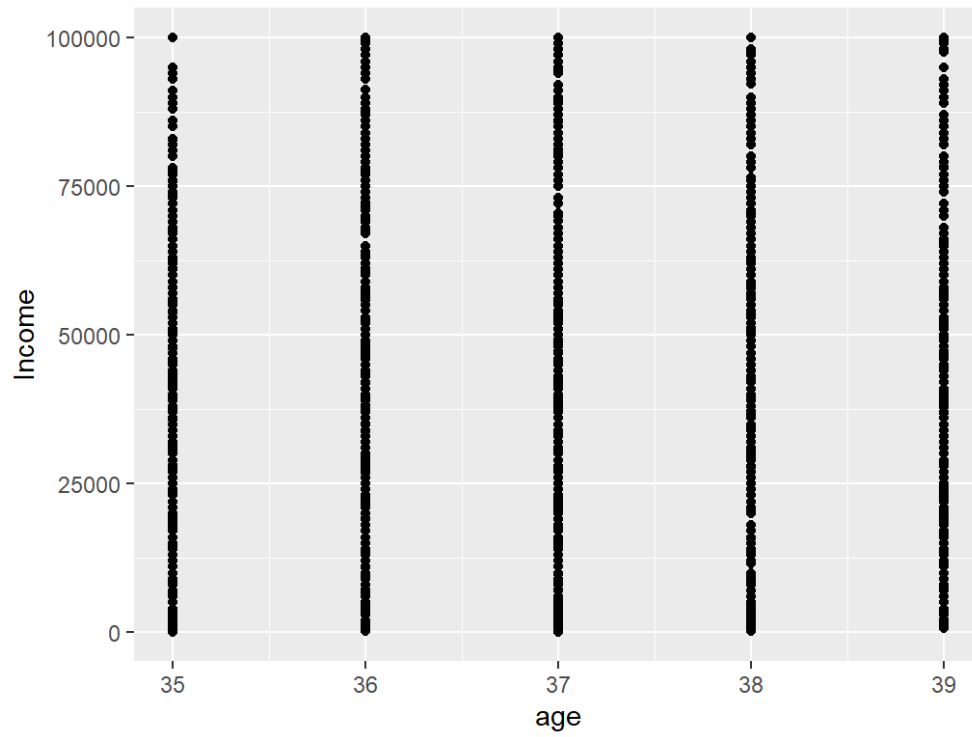
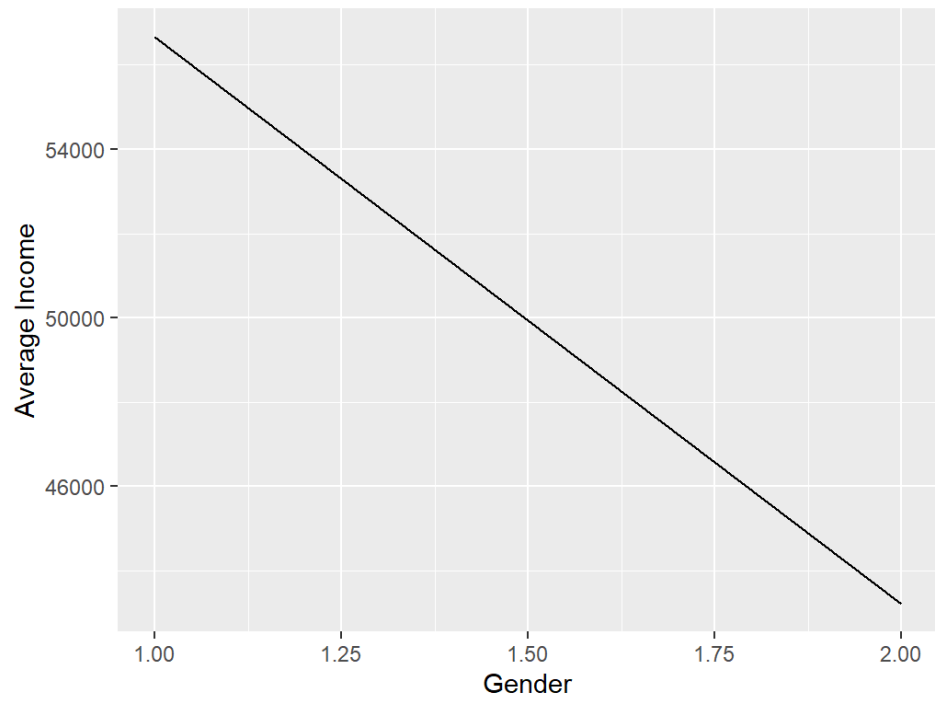
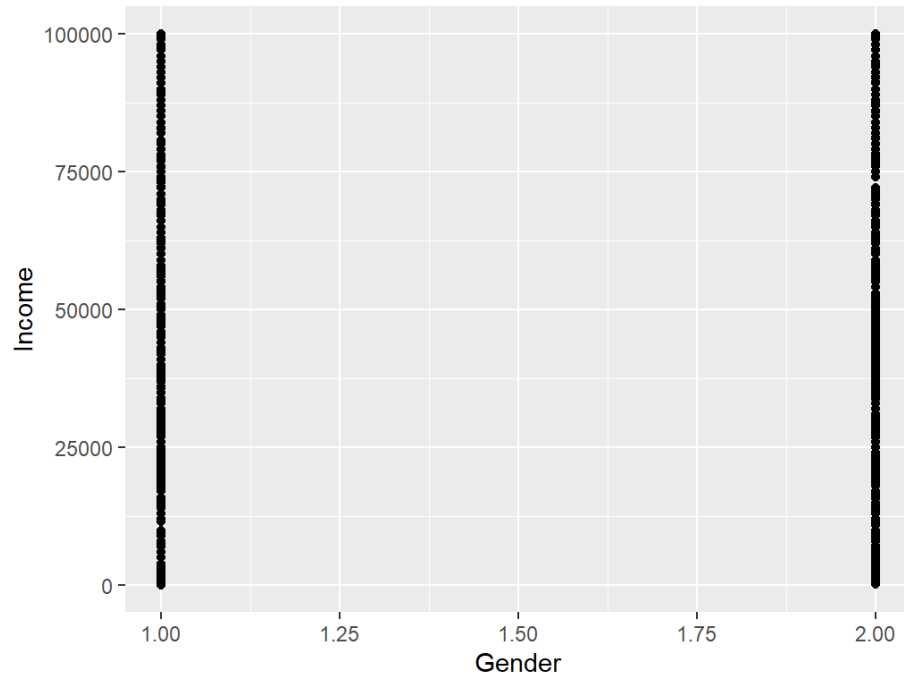
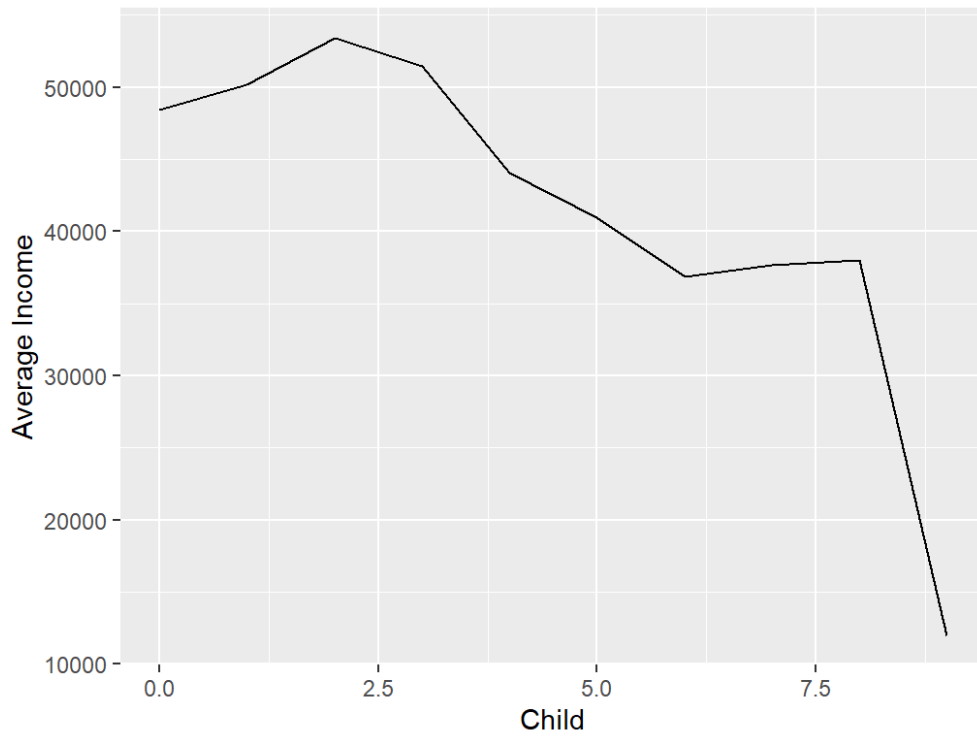
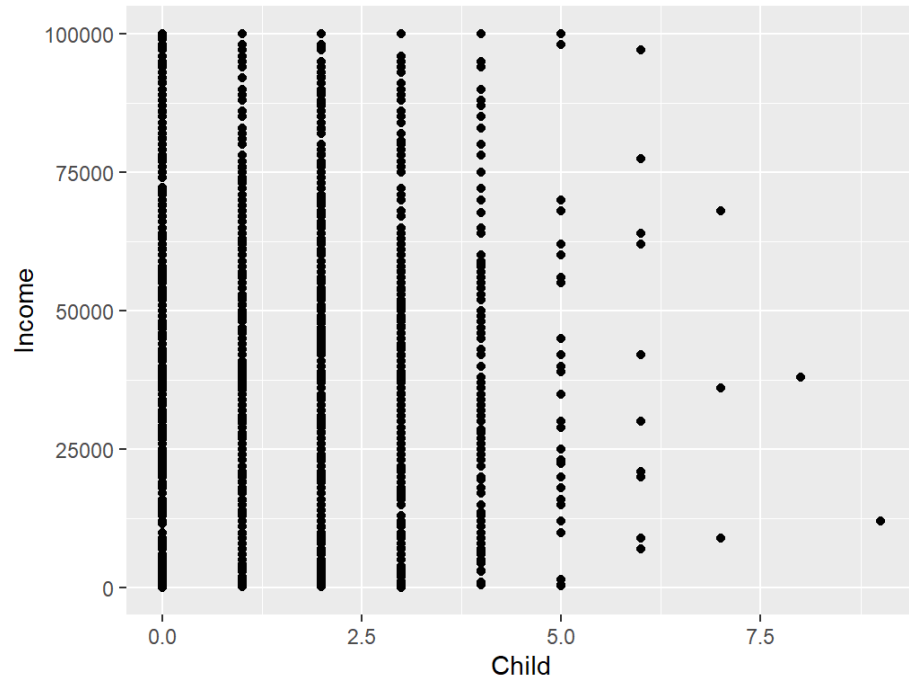


# Problem 1

(c)







	age	`YINC_1700_2019 == 0`	n
	<dbl>	<lgl>	<int>
1	35	FALSE	<u>1066</u>
2	35	TRUE	705
3	36	FALSE	<u>1104</u>
4	36	TRUE	703
5	37	FALSE	<u>1101</u>
6	37	TRUE	740
7	38	FALSE	<u>1106</u>
8	38	TRUE	768
9	39	FALSE	999
10	39	TRUE	692

	KEY_SEX_1997	`YINC_1700_2019 == 0`	n
	<dbl>	<lgl>	<int>
1	1	FALSE	<u>2779</u>
2	1	TRUE	<u>1820</u>
3	2	FALSE	<u>2597</u>
4	2	TRUE	<u>1788</u>

	CV_BIO_CHILD_HH_U18_2019	`YINC_1700_2019 == 0`	n
	<dbl>	<lgl>	<int>
1	0	FALSE	<u>1988</u>
2	0	TRUE	2696
3	1	FALSE	<u>1138</u>
4	1	TRUE	266
5	2	FALSE	<u>1385</u>
6	2	TRUE	343
7	3	FALSE	618
8	3	TRUE	183
9	4	FALSE	183
10	4	TRUE	82
11	5	FALSE	45
12	5	TRUE	25
13	6	FALSE	14
14	6	TRUE	10
15	7	FALSE	3
16	7	TRUE	2
17	8	FALSE	1
18	8	TRUE	1
19	9	FALSE	1

> |

## Problem 2:

(a)

```
Call:
lm(formula = log(income) ~ Gender + biodad + biomom + resdad +
    resmom + hhsize + marstat + child + urbanrural + yearssch +
    age + work_exp + black + hispanic + nonbnonh, data = posinc)
```

Residuals:

Min	1Q	Median	3Q	Max
-5.5486	-0.2566	0.1248	0.4510	1.8808

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	8.8686286	0.2980580	29.755	< 2e-16	***
Gender	-0.4129094	0.0211414	-19.531	< 2e-16	***
biodad	0.0006353	0.0022445	0.283	0.77714	
biomom	0.0030392	0.0037020	0.821	0.41170	
resdad	0.0028173	0.0018457	1.526	0.12696	
resmom	0.0032969	0.0032120	1.026	0.30473	
hhsize	-0.0524151	0.0098494	-5.322	1.07e-07	***
marstat	0.0391630	0.0110955	3.530	0.00042	***
child	0.0695484	0.0127932	5.436	5.68e-08	***
urbanrural	0.1346791	0.0254952	5.283	1.32e-07	***
yearssch	0.0979461	0.0038117	25.696	< 2e-16	***
age	0.0048824	0.0073897	0.661	0.50883	
work_exp	0.0337307	0.0019249	17.523	< 2e-16	***
black	-0.1095982	0.1104869	-0.992	0.32126	
hispanic	0.0460811	0.1106574	0.416	0.67711	
nonbnonh	0.0069920	0.1095313	0.064	0.94910	

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.7494 on 5360 degrees of freedom  
Multiple R-squared: 0.2437, Adjusted R-squared: 0.2416  
F-statistic: 115.1 on 15 and 5360 DF, p-value: < 2.2e-16

(c)

Call:

```
glm(formula = inlf ~ Gender + biodad + biomom + resdad + resmom +  
    hhsize + marstat + child + urbanrural + yearssch + age +  
    work_exp + black + hispanic + nonbnonh, family = binomial(link = "probi  
t"),  
    data = dat_A4)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-3.9930	0.1111	0.4499	0.7380	1.6437

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )	
(Intercept)	-0.6934906	0.5308406	-1.306	0.191416	
Gender	-0.2849209	0.0386732	-7.367	1.74e-13	***
biodad	0.0132585	0.0038303	3.462	0.000537	***
biomom	0.0001498	0.0058949	0.025	0.979730	
resdad	-0.0051748	0.0033518	-1.544	0.122616	
resmom	-0.0012037	0.0052882	-0.228	0.819934	
hhsize	-0.0050173	0.0162398	-0.309	0.757359	
marstat	0.0798707	0.0193745	4.122	3.75e-05	***
child	0.0205641	0.0213982	0.961	0.336542	
urbanrural	0.0276423	0.0452841	0.610	0.541583	
yearssch	0.0851288	0.0068540	12.420	< 2e-16	***
age	-0.0074769	0.0131342	-0.569	0.569175	
work_exp	0.1114958	0.0047744	23.353	< 2e-16	***
black	-0.0747970	0.1965287	-0.381	0.703507	
hispanic	0.1219491	0.1977375	0.617	0.537418	
nonbnonh	0.1067581	0.1958305	0.545	0.585647	

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 7427.3 on 6946 degrees of freedom  
Residual deviance: 6069.6 on 6931 degrees of freedom  
AIC: 6101.6

This is the estimation of the probability of a person participate in the labor market.

Second Stage:

Call:

```
lm(formula = log(income) ~ IMR + Gender + biodad + biomom + resdad +  
    resmom + hhsiz + marstat + child + urbanrural + yearssch +  
    age + work_exp + black + hispanic + nonbnh, data = posinc)
```

Residuals:

Min	1Q	Median	3Q	Max
-5.4448	-0.2530	0.1236	0.4332	2.4475

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	10.528215	0.321893	32.707	< 2e-16	***
IMR	-1.324523	0.105036	-12.610	< 2e-16	***
Gender	-0.268609	0.023772	-11.299	< 2e-16	***
biodad	-0.003535	0.002237	-1.580	0.114087	
biomom	0.001310	0.003651	0.359	0.719780	
resdad	0.006714	0.001845	3.639	0.000276	***
resmom	0.006381	0.003175	2.010	0.044517	*
hhsiz	-0.086343	0.010073	-8.571	< 2e-16	***
marstat	-0.016917	0.011805	-1.433	0.151922	
child	0.084742	0.012666	6.690	2.45e-11	***
urbanrural	0.049103	0.026028	1.887	0.059271	.
yearssch	0.028903	0.006640	4.353	1.37e-05	***
age	0.009754	0.007293	1.337	0.181170	
work_exp	-0.007907	0.003808	-2.076	0.037912	*
black	-0.094789	0.108900	-0.870	0.384108	
hispanic	0.029241	0.109070	0.268	0.788635	
nonbnh	-0.037388	0.108009	-0.346	0.729238	

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

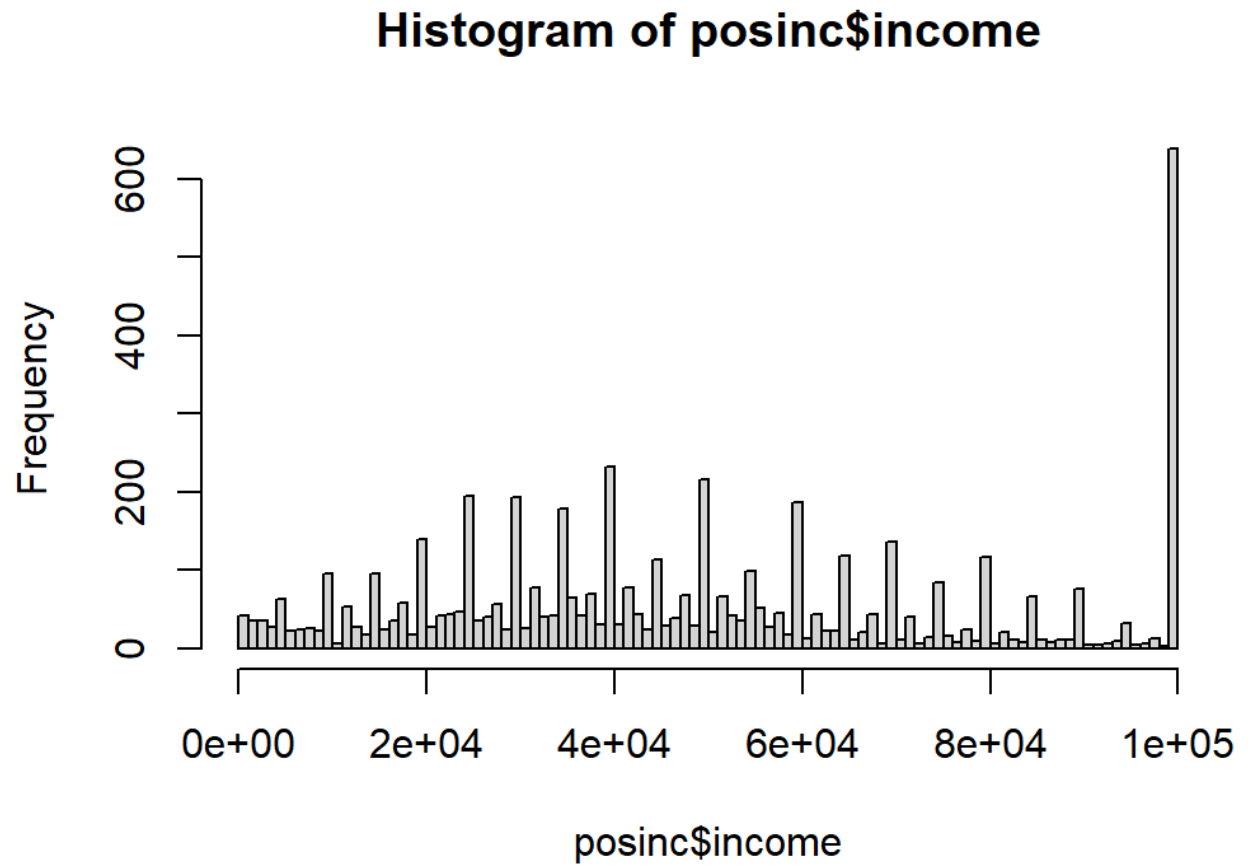
Residual standard error: 0.7386 on 5359 degrees of freedom

Multiple R-squared: 0.2655, Adjusted R-squared: 0.2633

F-statistic: 121.1 on 16 and 5359 DF, p-value: < 2.2e-16

### Problem 3:

(a)



100,000 is the top coded value.



(b,c,d)

	Estimates
Gender	-0.484340492
biodad	0.001362190
biomom	0.003386790
resdad	0.004032734
resmom	0.007227052
hhsiz	-0.031851519
marstat	0.032390773
child	0.068149624
urbanrural	0.220244260
yearssch	0.131366171
age	0.200516794
work_exp	0.032688270
black	0.954175976
hispanic	1.129718251
nonbnh	1.092358518

This is the estimates after correcting the censoring problem.

## Problem 4

Between estimator:

Call:

```
lm(formula = Y ~ educ + exp + mstat, data = within)
```

Residuals:

Min	1Q	Median	3Q	Max
-36037	-5421	-1093	3214	85446

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-20335.65	605.34	-33.59	<2e-16 ***
educ	2493.81	59.38	42.00	<2e-16 ***
exp	2430.18	65.18	37.28	<2e-16 ***
mstat	6297.16	470.08	13.40	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 9917 on 8980 degrees of freedom

Multiple R-squared: 0.417, Adjusted R-squared: 0.4168

F-statistic: 2141 on 3 and 8980 DF, p-value: < 2.2e-16

#### Within Estimator:

```
Call:
lm(formula = Ywithin ~ eduwithinols + expwithinols + mstatwithinols,
    data = within)

Residuals:
    Min       1Q   Median       3Q      Max
-105335   -7105        0    5741   303070

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  -1.934e-10  4.223e+01   0.00    1
eduwithinols   2.798e+03  2.394e+01  116.88 <2e-16 ***
expwithinols   2.632e+03  1.826e+01  144.15 <2e-16 ***
mstatwithinols 1.093e+04  1.485e+02   73.61 <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 16980 on 161708 degrees of freedom
Multiple R-squared:  0.349,    Adjusted R-squared:  0.349
F-statistic: 2.89e+04 on 3 and 161708 DF,  p-value: < 2.2e-16
```

#### Difference Estimator:

```
Call:
lm(formula = differenceinc ~ differenceedu + differenceexp +
    differencemstat, data = Difference)

Residuals:
    Min       1Q   Median       3Q      Max
-236160   -3780    -998    2661   344130

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    997.68     42.92   23.24 <2e-16 ***
differenceedu  1896.02     28.85   65.72 <2e-16 ***
differenceexp  1787.98     21.50   83.16 <2e-16 ***
differencemstat 3527.52    115.12   30.64 <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 16550 on 152724 degrees of freedom
Multiple R-squared:  0.1253,    Adjusted R-squared:  0.1253
F-statistic: 7293 on 3 and 152724 DF,  p-value: < 2.2e-16
```