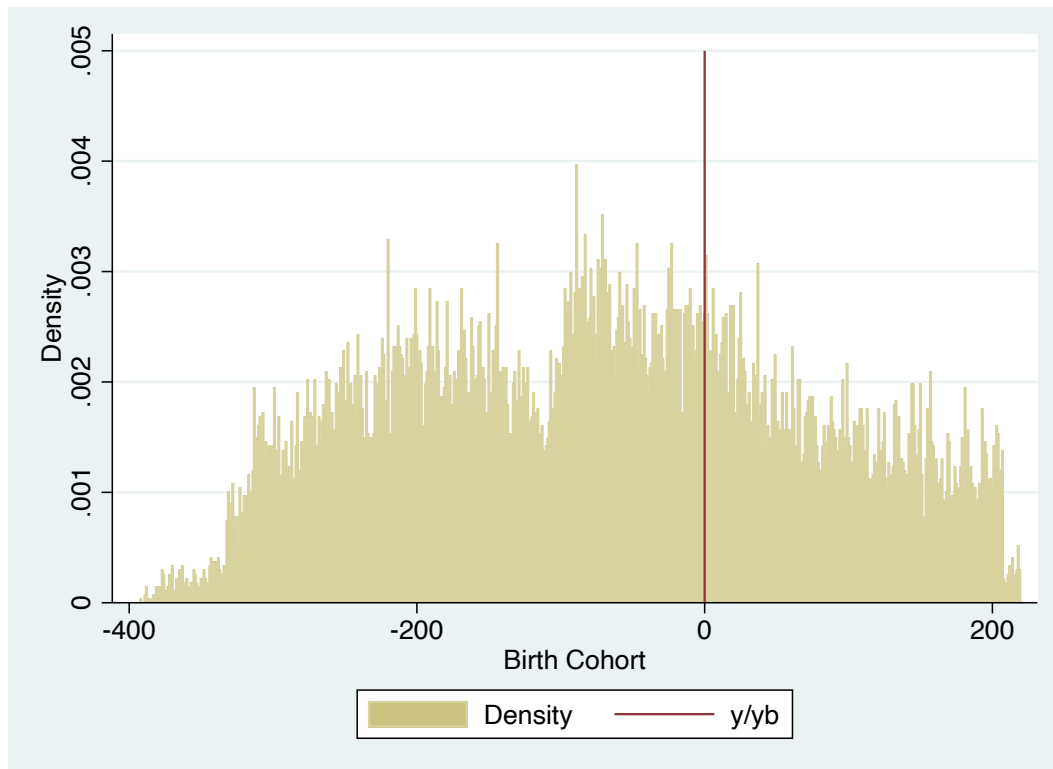


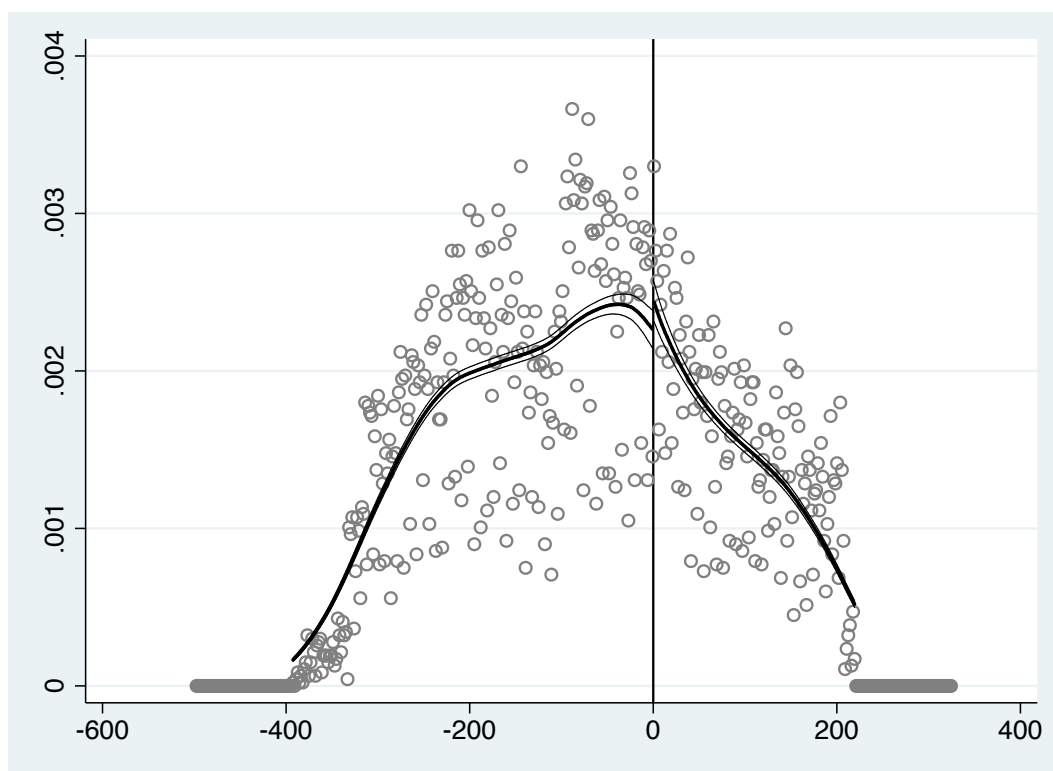
ECO 3211 PS4  
119020339  
Ouyang Yingrun

### 3 Validity test pf RDD

#### b. Density test of x: histogram



#### Density test: DCdensity



## Density test of x: regression

//First stage

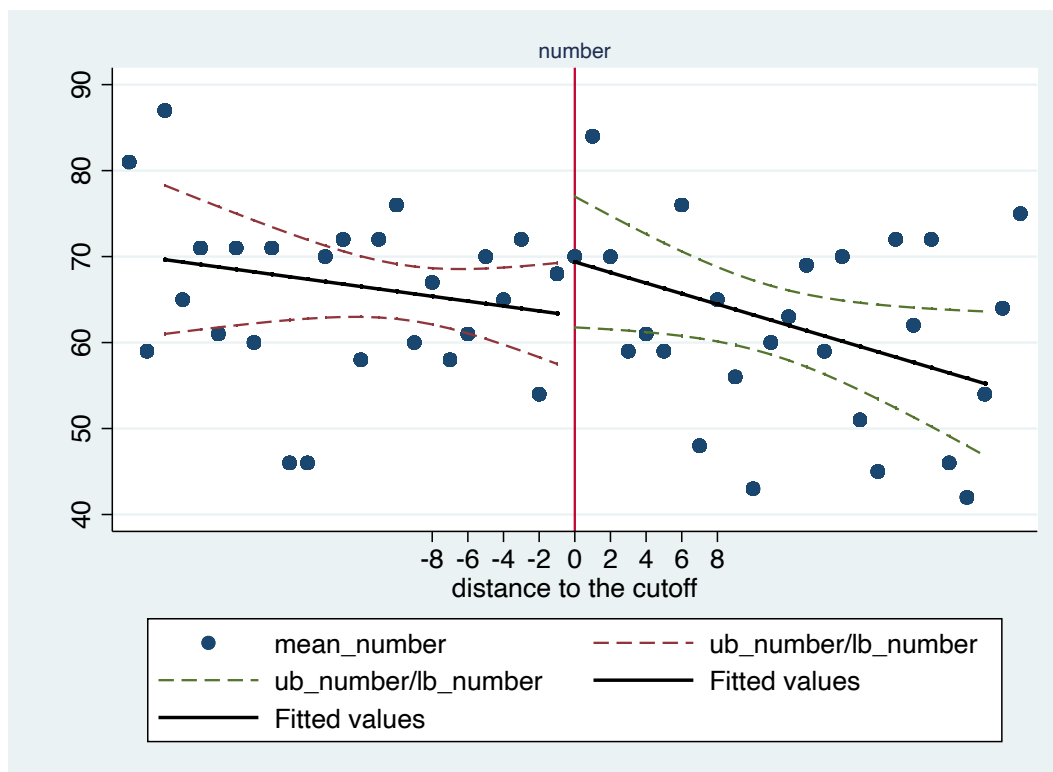
```
. reg number d x dx if abs(x)<=bw_reduce_number_CCT, cluster(x) //rectangle kernal weight
```

Linear regression	Number of obs	=	2,957
	F(3, 46)	=	2.08
	Prob > F	=	0.1162
	R-squared	=	0.1460
	Root MSE	=	9.2997

(Std. Err. adjusted for 47 clusters in x)

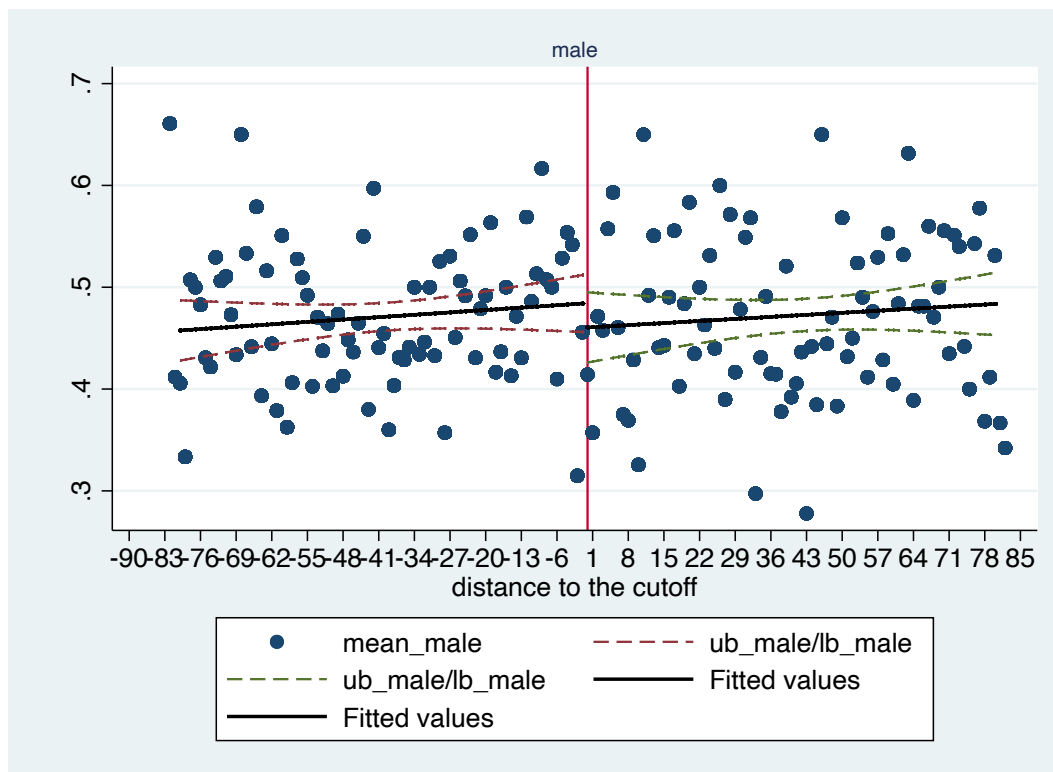
number	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
d	6.270595	5.064493	1.24	0.222	-3.9237	16.46489
x	-.2843127	.3010447	-0.94	0.350	-.8902844	.3216589
dx	-.3304623	.430194	-0.77	0.446	-1.196398	.5354734
_cons	63.10552	3.251581	19.41	0.000	56.56043	69.65062

## Density regression

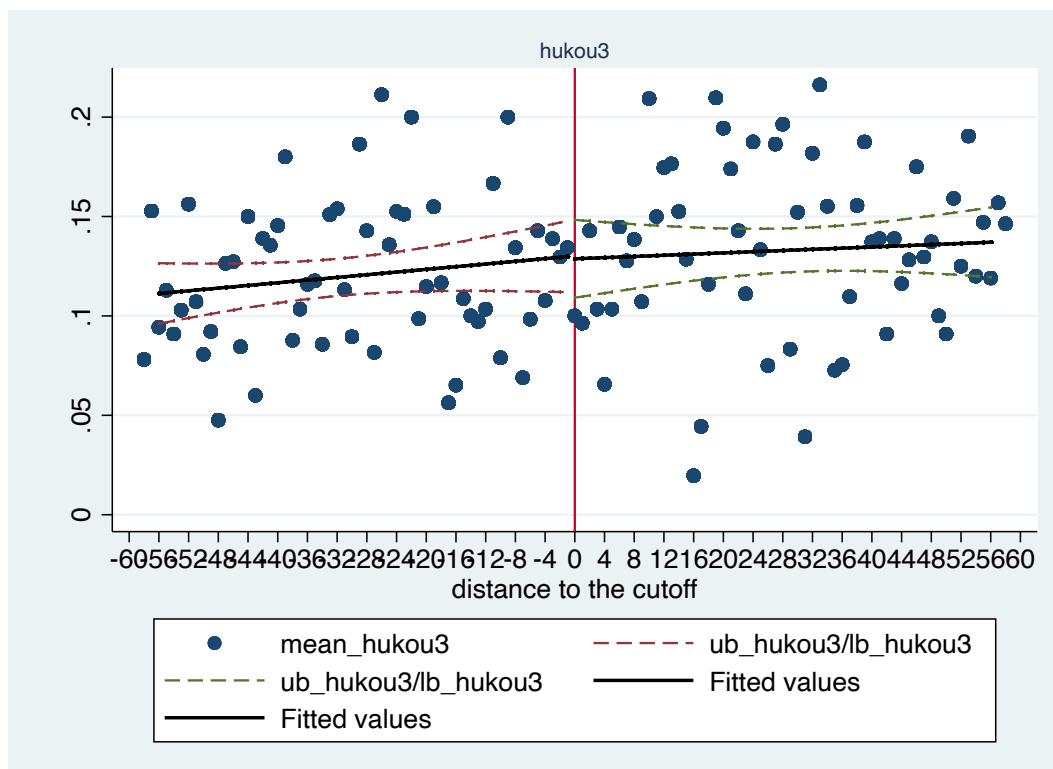


C. smoothness of predetermined characteristics: all predetermined characteristics showing a smoothness outcome except for father's education.

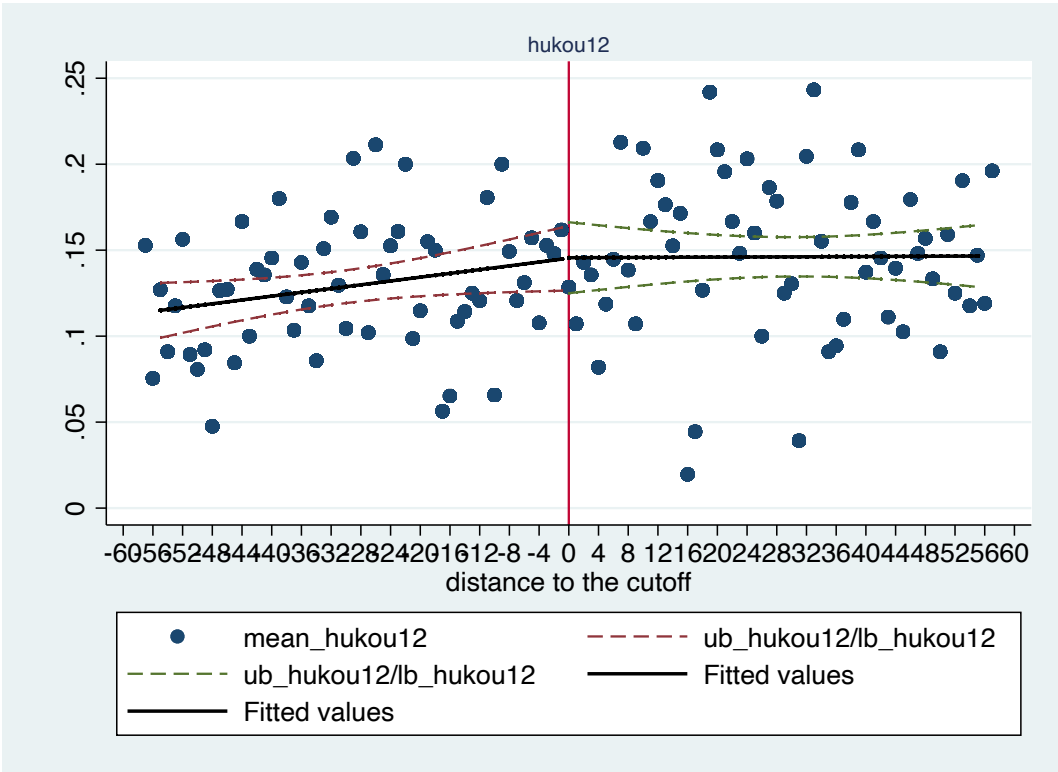
Gender



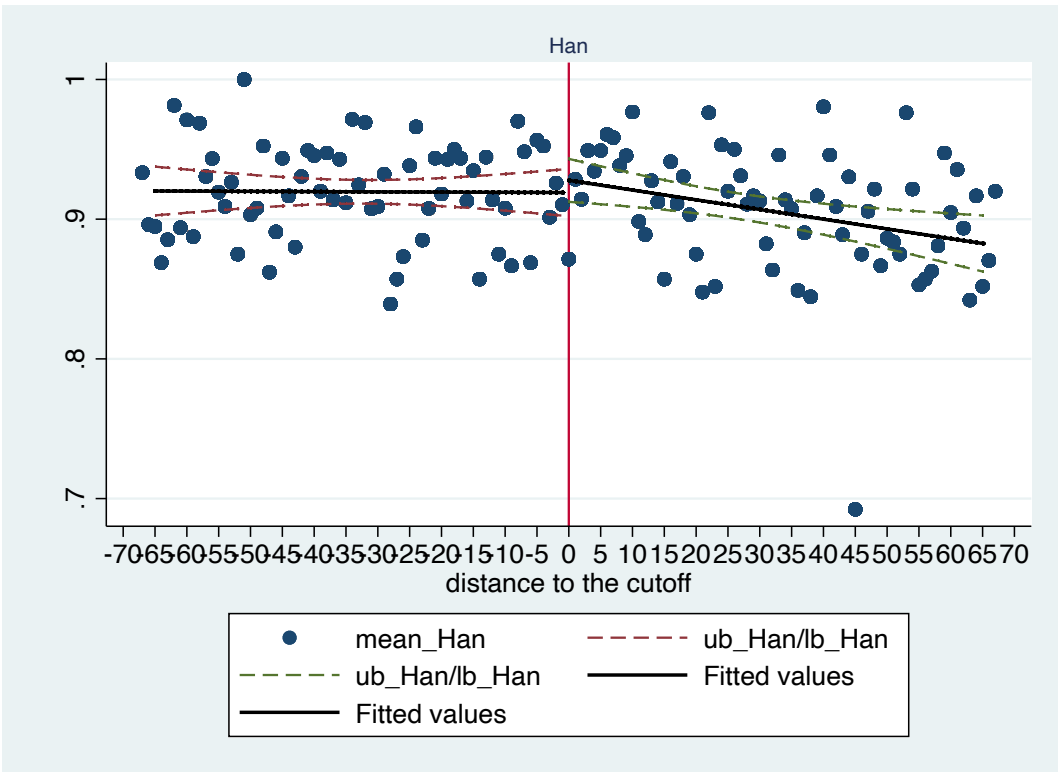
Whether had an urban Hukou at age 3



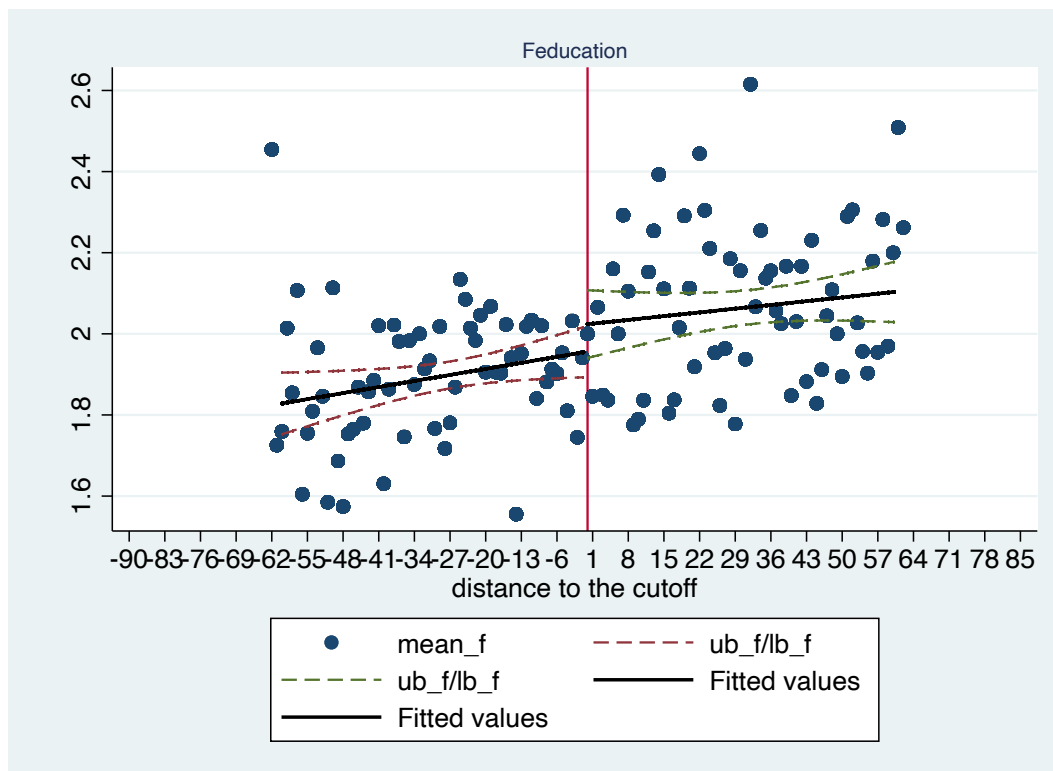
Whether had an urban Hukou at age 12



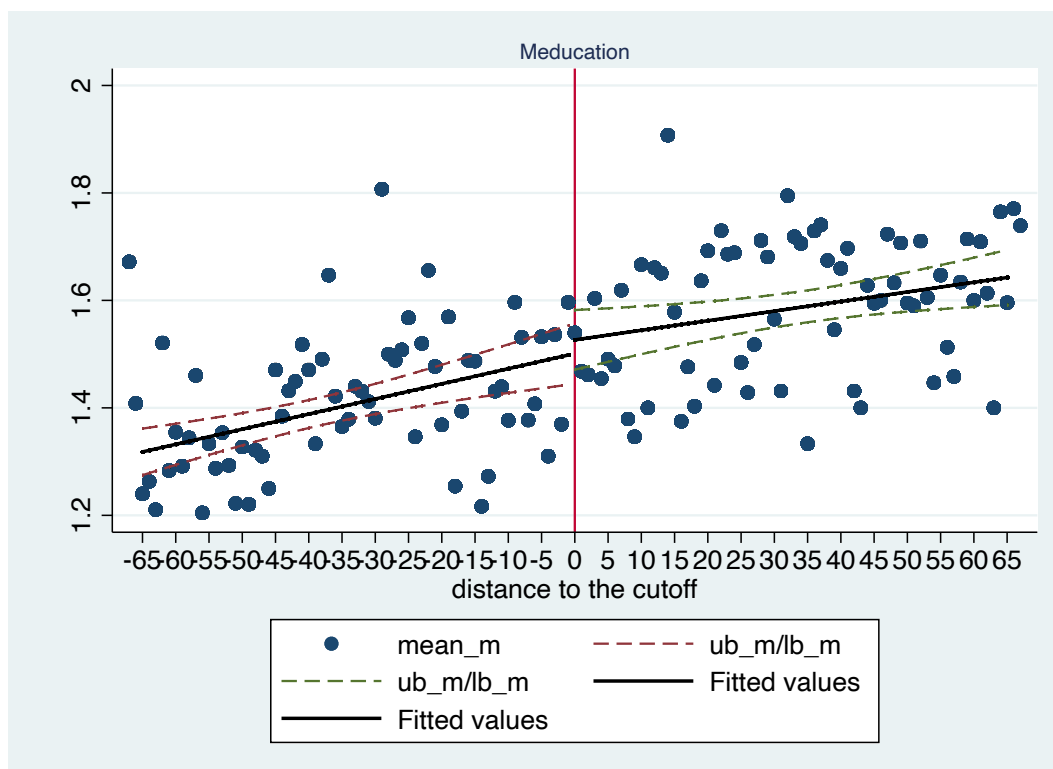
Han groups



## Father's education



## Mother's education



#### 4. outcomes of nonparametric approach:

We mainly discuss the estimated coefficient of  $d$  which captures the impact of compulsory education on total years of schooling. Our results shown that compulsory education have a significant effect on years of schooling. Being covered by compulsory education would increase 0.455 years of schooling, and outcome is statistically significant at 5%.

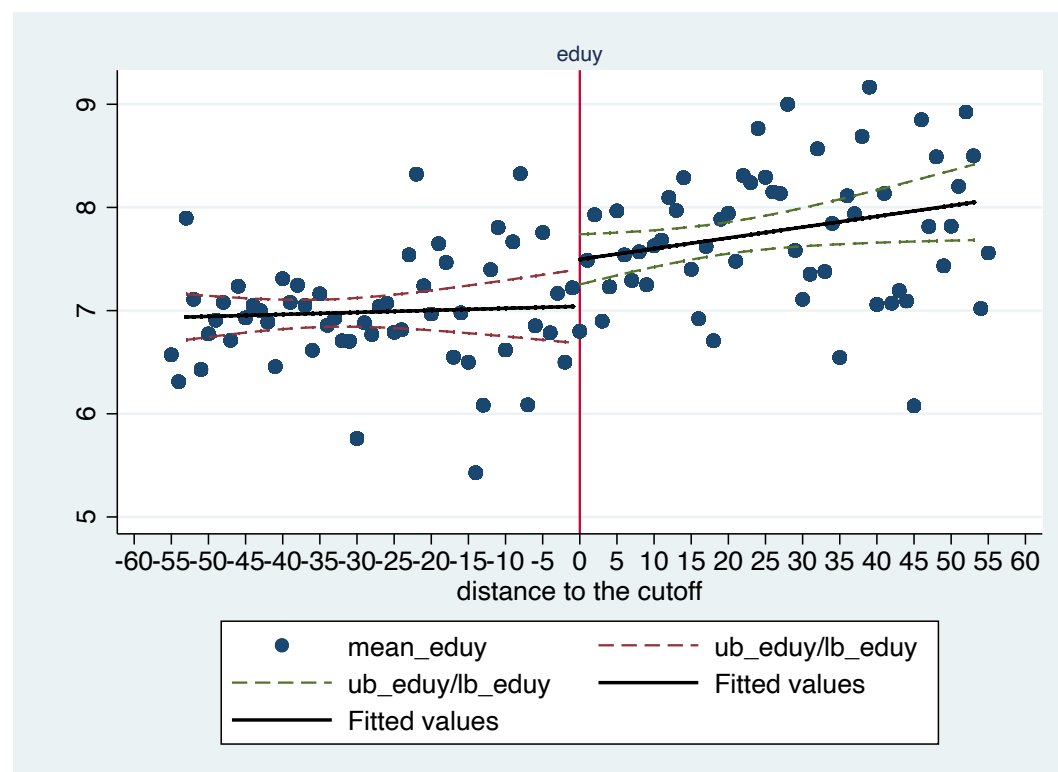
```
. *reduce form
. reg eduy d x dx if abs(x)<=bw_reduce_eduy_CCT, cluster(x) //rectangle kernal weight
```

Linear regression

Number of obs	=	6,339
F(3, 106)	=	17.81
Prob > F	=	0.0000
R-squared	=	0.0084
Root MSE	=	4.313

(Std. Err. adjusted for 107 clusters in x)

eduy	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
d	.4548502	.2214598	2.05	0.042	.0157846	.8939157
x	.0019569	.0049329	0.40	0.692	-.007823	.0117368
dx	.0084669	.0070344	1.20	0.231	-.0054794	.0224132
_cons	7.041691	.1834547	38.38	0.000	6.677974	7.405408



## 5. Check the robustness of your findings by varying the bandwidth choices

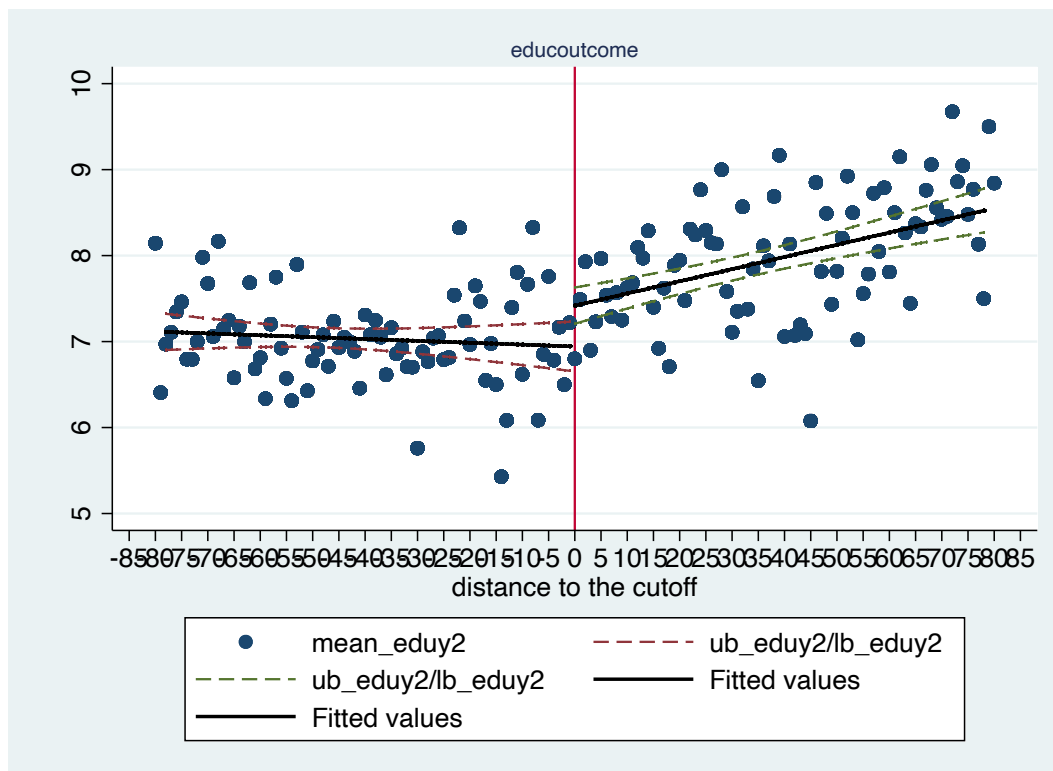
```
. reg eduy d x dx if abs(x)<=bw_reduce_eduy_IK, cluster(x) //rectangle kernal weight
```

Linear regression

Number of obs	=	9,165
F(3, 156)	=	41.09
Prob > F	=	0.0000
R-squared	=	0.0131
Root MSE	=	4.3008

(Std. Err. adjusted for 157 clusters in x)

eduy	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]
d	.4750621	.1829498	2.60	0.010	.1136837 .8364405
x	-.0021684	.0029036	-0.75	0.456	-.0079039 .0035671
dx	.0163413	.0038238	4.27	0.000	.0087882 .0238944
_cons	6.942969	.1479574	46.93	0.000	6.650711 7.235228



```
. // Choose CV
. reg eduy d x dx if abs(x)<=bw_reduce_eduy_CV, cluster(x) //rectangle kernal weight
```

Linear regression

Number of obs	=	9,058
F(3, 154)	=	43.45
Prob > F	=	0.0000
R-squared	=	0.0134
Root MSE	=	4.296

(Std. Err. adjusted for 155 clusters in x)

eduy	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]
d	.4601403	.1835669	2.51	0.013	.097506 .8227745
x	-.0023134	.0029917	-0.77	0.441	-.0082235 .0035968
dx	.0172928	.0038463	4.50	0.000	.0096945 .0248912
_cons	6.939137	.1496101	46.38	0.000	6.643584 7.23469

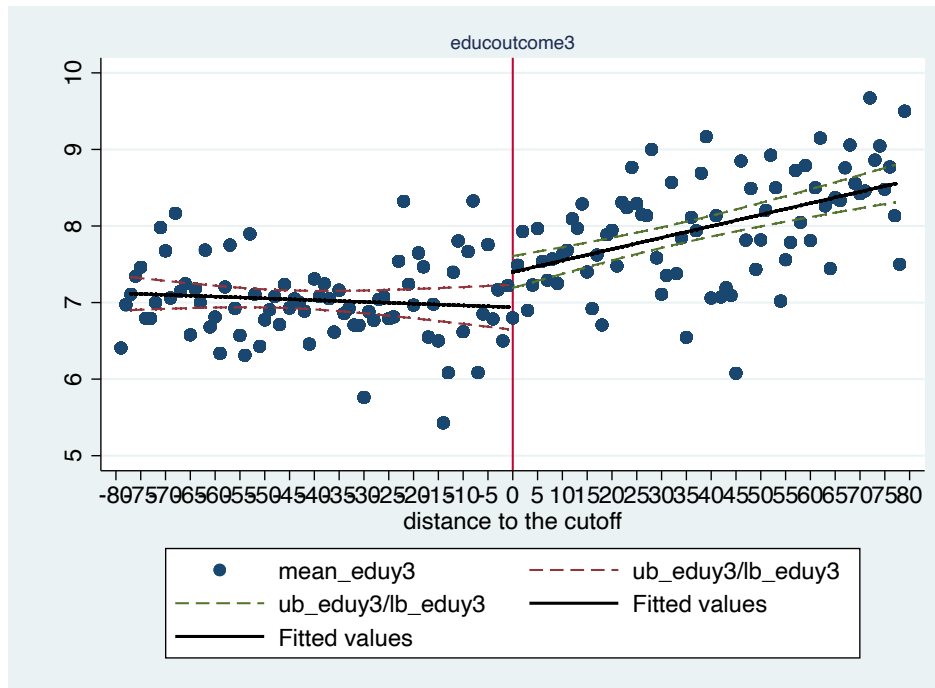
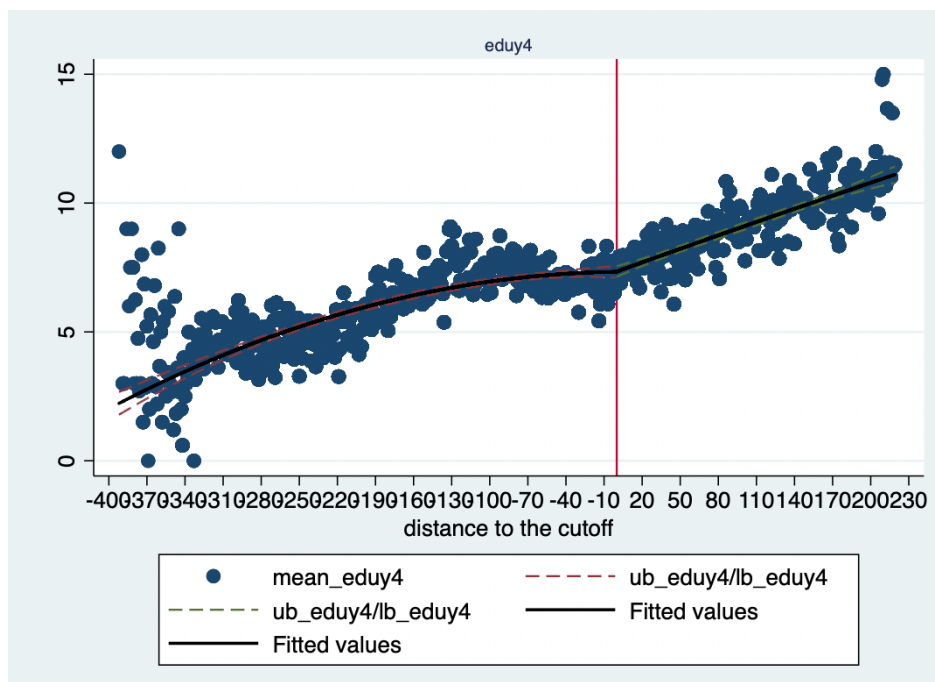


Table presents the outcome of varying bandwidth choice. We find that overall outcome of compulsory education impacts is consistent with the outcome of estimation equation shown in previous table.

Check the robustness of your findings by using the parametric estimation approach.





```
. reg eduy d x dx x2 dx2, cluster(x)
```

```
Linear regression      Number of obs   =    26,694
                      F(5, 607)         =    670.03
                      Prob > F           =    0.0000
                      R-squared          =    0.1270
                      Root MSE         =    4.3696
```

(Std. Err. adjusted for 608 clusters in x)

eduy	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
d	.0084778	.1541863	0.05	0.956	-.2943255	.3112811
x	.0006577	.0016607	0.40	0.692	-.0026037	.0039191
dx	.0171597	.0030164	5.69	0.000	.0112358	.0230837
x2	-.0000316	4.83e-06	-6.53	0.000	-.000041	-.0000221
dx2	.0000286	.0000132	2.16	0.031	2.59e-06	.0000546
_cons	7.33394	.1123944	65.25	0.000	7.113211	7.554669

Table presents the outcome of parametric approach regression. By using a parametric regression, we find that overall outcome of compulsory education impacts is economically smaller comparing with the outcome of estimation equation shown in previous table. Namely, the outcome is not quite consistent with non-parametric approach.