

# RCT Implementation

## Randomized Control Trial

In this exercise, I'll provide some simple examples of how to implement a basic RCT analysis.

Before we start, however, we need to create some data.

1. Estimate two wages models. For men and women, using an heteroskedastic regression model.

The goal: estimate two potential outcomes for wages. One following Womens wage structure, and the other using Men wage structure.

```
set linesize 255
frause oxaca, clear
set seed 101
qui:hetreg lnwage age agesq married divorced kids6 kids714 if female==0, het(age agesq mar
est sto m1
predict xb1, xb
predict s1, sigma
qui:hetreg lnwage age agesq married divorced kids6 kids714 if female==1, het(age agesq mar
est sto m2
predict xb0, xb
predict s0, sigma
```

(Excerpt from the Swiss Labor Market Survey 1998)

Here, as I did for UQregression, I obtain the predicted mean and predicted standard errors. and simulate two wages. I will use wage differences between simulated men and women predicted log wages as the treated effect

$$TE = \ln \hat{w}_1 - \ln \hat{w}_0$$

$$\ln \hat{w}_1 \sim N(xb_1, s_1^2); \ln \hat{w}_0 \sim N(xb_0, s_0^2)$$

```
gen lnwage1 = rnormal(xb0,s0)
gen lnwage0 = rnormal(xb1,s1)
gen teff=lnwage1-lnwage0
```

I will create a randomized treatment, and assume those treated get the treatment effect

$$Y_i = Y_i(0) + trt_i * \delta_i$$

```
gen trt = runiform()<.5
replace lnwage = lnwage0 + trt * teff
```

(1,647 real changes made)

## Visual Exploration

Now that we have a randomized treatment, we could start exploring the data:

```
two (kdensity lnwage if trt == 1) (kdensity lnwage if trt == 0) , ///
    legend(order(1 "Treated" 2 "Untreated"))
```

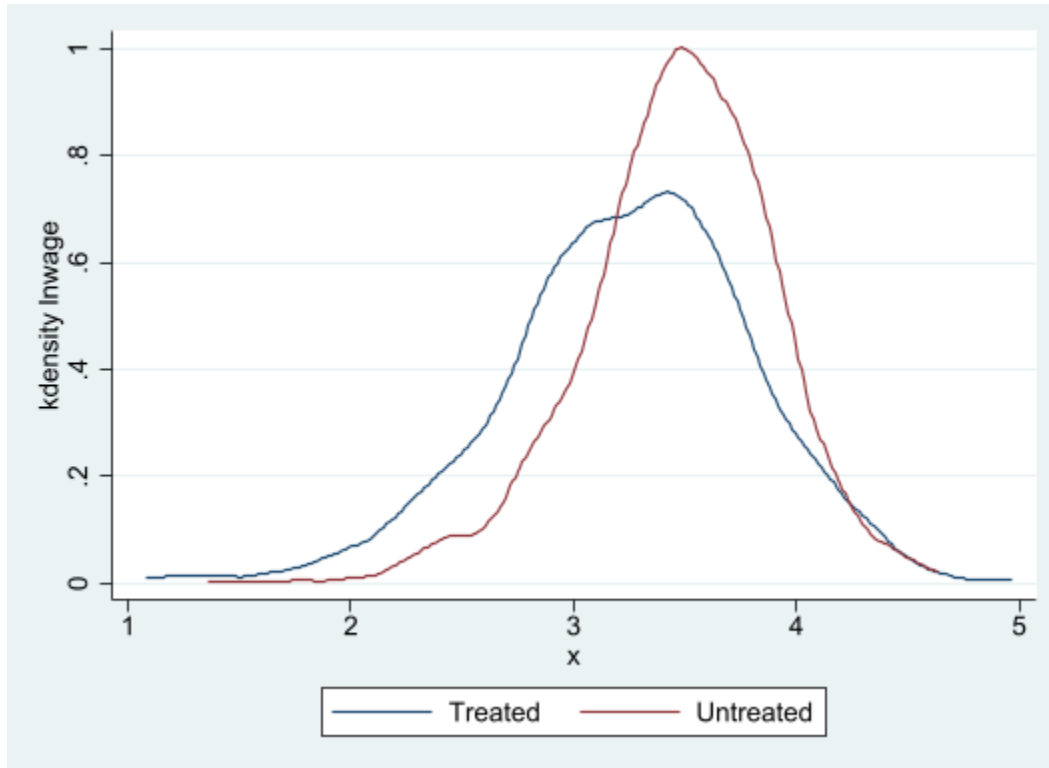


Figure 1: Log wage distribution between Treated and untreated

In order to estimate the treatment effects, we could simple estimate a regression model of the outcome. Compare it to the treatment effect

### Estimation of ATE Effect

```
** True Effect
sum teff
** Simple Regression
set linesize 255
reg lnwage trt, robust
est sto m0
```

Variable	Obs	Mean	Std. dev.	Min	Max
teff	1,647	-.2121816	.6613419	-3.024343	2.704082

Linear regression	Number of obs	=	1,647
	F(1, 1645)	=	79.85
	Prob > F	=	0.0000
	R-squared	=	0.0464
	Root MSE	=	.5017

		Robust					
lnwage		Coefficient	std. err.	t	P> t	[95% conf. interval]	
-----+-----							
trt							
		-.2213292	.0247679	-8.94	0.000	-.2699092	-.1727492
_cons		3.465982	.0150123	230.88	0.000	3.436537	3.495427

Because treatment is randomized, we could also add other controls to the model, and improve on precision

```

qui:reg lnwage trt age agesq , robust
est sto m1
qui:reg lnwage trt age agesq married divorced , robust
est sto m2
qui:reg lnwage trt age agesq married divorced kids6 kids714 , robust
est sto m3

esttab m0 m1 m2 m3, se nonum mtitle("m0" "m1" "m2" "m3") keep(trt) md

```

	m0	m1	m2	m3
trt	-0.221*** (0.0248)	-0.214*** (0.0225)	-0.213*** (0.0225)	-0.212*** (0.0224)
N	1647	1647	1647	1647

Standard errors in parentheses \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

## Falsification

We could just use other outcomes that shouldn't be affected by the treatment. You expect they have no impact on outcome

```
qui:reg exper trt age agesq married divorced kids6 kids714 , robust
est sto m0
qui:reg tenure trt age agesq married divorced kids6 kids714 , robust
est sto m1
esttab m0 m1 , se nonum mtitle("m0" "m1") keep(trt) md
```

	m0	m1
trt	-0.104 (0.366)	-0.458 (0.335)
N	1434	1434

Standard errors in parentheses \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

## Balance test

You should also try to create balance tables, where you compare and test if characteristics are similar across treated and control groups:

```
tabstat age agesq married divorced kids6 kids714 , by(trt)
sureg age agesq married divorced kids6 kids714 =trt,
```

Summary statistics: Mean

Group variable: trt

trt	age	agesq	married	divorced	kids6	kids714
0	39.14475	1649.63	.53076	.1206273	.2979493	.3365501
1	39.3643	1675.521	.5158924	.1466993	.2713936	.3215159
Total	39.25379	1662.489	.5233758	.1335762	.2847602	.3290832

# Seemingly unrelated regression

Equation	Obs	Params	RMSE	"R-squared"	chi2	P>chi2
age	1,647	1	11.02798	0.0001	0.16	0.6862
agesq	1,647	1	893.7224	0.0002	0.35	0.5566
married	1,647	1	.4993979	0.0002	0.36	0.5458
divorced	1,647	1	.3399466	0.0015	2.42	0.1197
kids6	1,647	1	.6626276	0.0004	0.66	0.4161
kids714	1,647	1	.7071256	0.0001	0.19	0.6662

		Coefficient	Std. err.	z	P> z	[95% conf. interval]
age						
	trt	.2195505	.5434865	0.40	0.686	-.8456636 1.284765
	_cons	39.14475	.3830175	102.20	0.000	38.39405 39.89545
agesq						
	trt	25.89111	44.04489	0.59	0.557	-60.43528 112.2175
	_cons	1649.63	31.04026	53.14	0.000	1588.792 1710.467
married						
	trt	-.0148675	.0246116	-0.60	0.546	-.0631054 .0333703
	_cons	.53076	.0173448	30.60	0.000	.4967648 .5647552
divorced						
	trt	.026072	.0167534	1.56	0.120	-.0067641 .0589081
	_cons	.1206273	.0118068	10.22	0.000	.0974863 .1437682
kids6						
	trt	-.0265557	.032656	-0.81	0.416	-.0905602 .0374488
	_cons	.2979493	.023014	12.95	0.000	.2528427 .343056
kids714						
	trt	-.0150342	.0348489	-0.43	0.666	-.0833368 .0532685
	_cons	.3365501	.0245595	13.70	0.000	.2884143 .3846858

Here, the goal is just to see if `trt` is not-significant across groups