

Matching (cont.)

observed : y_i, D_i, x_i
for i

$$y_i = D_i y_i(1) + (1 - D_i) y_i(0) \quad = \text{observed outcome for } i$$

Assumptions to estimate Δ^{ATE}

$$y(0), y(1) \perp D \mid x$$

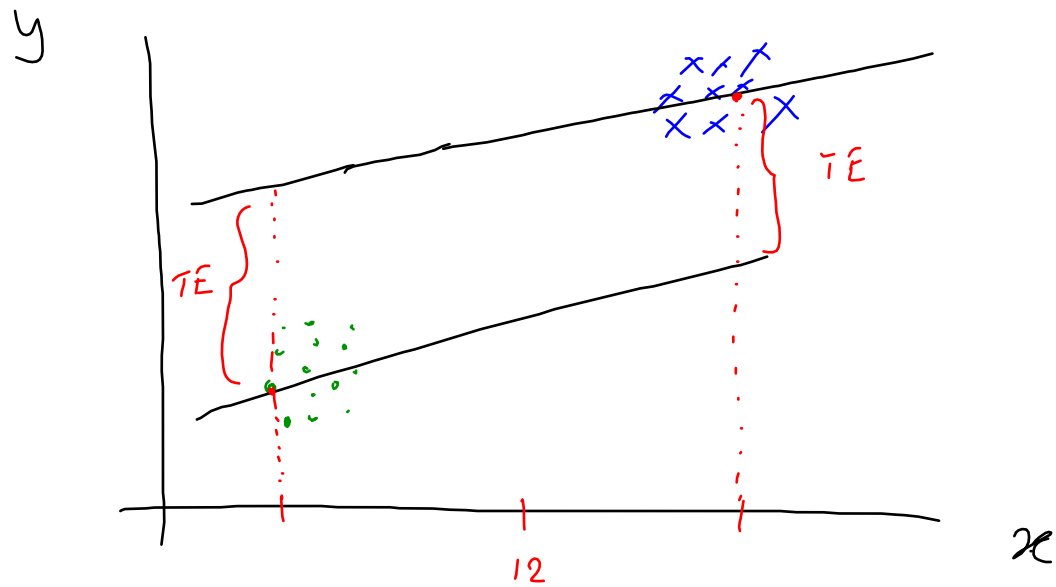
$\hookrightarrow$ independent
of

$$0 < P(D=1 \mid x) < 1$$

Random sample $\{y, D, x\}$

Matching $\rightarrow$ when comparing treated & control obs. more weight given to similar obs.

No functional form assumptions.



$$\hat{\Delta}^{ATE} = \frac{1}{N} \sum_i [\gamma_i(1) - \gamma_i(0)]$$

↓
infeasible → missing data / counterfactual

$\gamma_i(1)$ observed if $D_i = 1$; $\gamma_i(0)$ unobserved if $D_i = 1$

Replace $\gamma_i(0)$ with a close match to i

but with $D = 0$ (e.g. $\gamma_j(0)$ where $D_j = 0$)

Various matching approaches to determine what weight to attach to observations from the opposite group.

One approach relies on propensity score, $p(x)$:

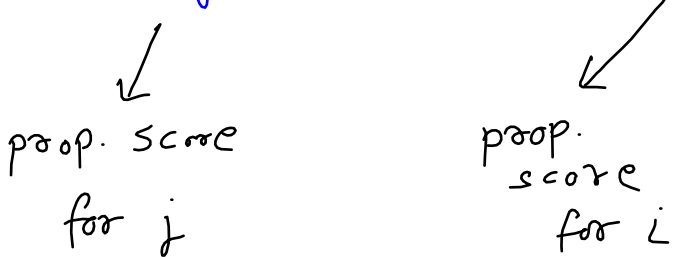
$$p(x) = \Pr(D=1 | x)$$

↳ typically estimated using logit / probit

$$\begin{array}{l} \gamma(0), \gamma(1) \perp D \mid x \\ \gamma(0), \gamma(1) \perp D \mid p(x) \end{array} \Rightarrow$$

Single nearest neighbor matching

If $D_i = 1$ & $Y_i(0)$ is unobserved use $Y_j(0) = y_j$
where $D_j = 0$ & p_j is closest to p_i



" $D_i = 0$ & $Y_i(1)$ use $Y_j(1) = y_j$
where $D_j = 1$ & p_j is closest to p_i

<u>obs.</u>	D	$Y(1)$	$Y(0)$	PS $P(D=1 X)$	Matched Y	TE
1	1	18	?	0.6 [✓]	12	6
2	1	24	?	0.7 [✓]	12	12
3	0	?	15	0.3 [✓]	16	1
4	0	?	12	0.4 ^{✓✓}	16	4
5	1	16	?	0.5 ^{✓✓}	12	4
						27

$$ATE = \frac{27}{5} = 5.4$$

Treated group

24 yrs

Family inc. = \$100,000

only child

high SAT

Control

D	age	family income	...
0			
0			
1			
1			
1			
0			