

Conditional Exchangeability

Sociol 114

Learning goals for today

At the end of class, you will be able to:

1. Define a conditionally randomized experiment
2. Define conditional exchangeability
3. Define conditional average treatment effects and recognize their use in policy

Review of exchangeability

Exchangeable sampling from a population

Population Outcomes	Randomized Sampling	Sampled Outcomes	Estimator:
Y_{Maria}	$S_{\text{Maria}} = 1$	Y_{Maria}	Estimate the population mean by the sample mean
Y_{William}	$S_{\text{William}} = 0$		
Y_{Rich}	$S_{\text{Rich}} = 0$		
Y_{Sarah}	$S_{\text{Sarah}} = 1$	Y_{Sarah}	
Y_{Alondra}	$S_{\text{Alondra}} = 0$		
$Y_{\text{Jesús}}$	$S_{\text{Jesús}} = 1$	$Y_{\text{Jesús}}$	

Key assumption:
Sampled and unsampled units are **exchangeable** due to random sampling

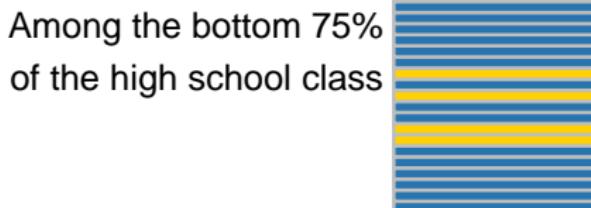
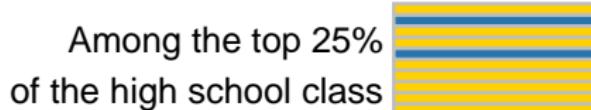
$$Y \perp S$$

Exchangeable treatment assignment

Population Potential Outcomes	Randomized Treatment	Observed Outcomes
Y^1_{Maria}	Y^0_{Maria}	$A_{\text{Maria}} = 1$
Y^1_{William}	Y^0_{William}	$A_{\text{William}} = 0$
Y^1_{Rich}	Y^0_{Rich}	$A_{\text{Rich}} = 0$
Y^1_{Sarah}	Y^0_{Sarah}	$A_{\text{Sarah}} = 1$
Y^1_{Alondra}	Y^0_{Alondra}	$A_{\text{Alondra}} = 0$
$Y^1_{\text{Jesús}}$	$Y^0_{\text{Jesús}}$	$A_{\text{Jesús}} = 1$

A **conditionally** randomized experiment

A hypothetical experiment: Conditional randomization



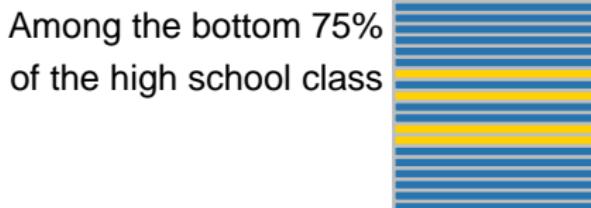
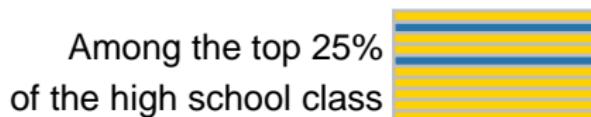
Randomly Assigned to

High School Degree
Four-Year College Degree

Outcome: Employed at age 40

Does exchangeability hold? How would you analyze?

A hypothetical experiment:
Conditional randomization



Randomly Assigned to

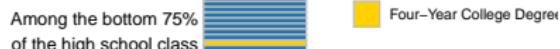
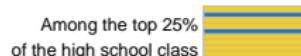
High School Degree
Four-Year College Degree

The legend consists of two colored squares: a blue square followed by the text "High School Degree" and a yellow square followed by the text "Four-Year College Degree".

Outcome: Employed at age 40

Conditional randomization: Exchangeability does not hold

A hypothetical experiment:
Conditional randomization



Randomly Assigned to

- High School Degree
- Four-Year College Degree

Conditional randomization: Exchangeability does not hold

Treated units are more likely to have done well in high school

A hypothetical experiment:
Conditional randomization

Among the top 25%
of the high school class



Among the bottom 75%
of the high school class



Randomly Assigned to
High School Degree
Four-Year College Degree

Conditional randomization: Exchangeability does not hold

Treated units are more likely to have done well in high school

Those who do well in high school are more likely to be employed at age 40 even without college

A hypothetical experiment:
Conditional randomization

Among the top 25%
of the high school class



The bar is composed of four horizontal segments: three yellow and one blue. The blue segment is approximately one-third of the total height.

Randomly Assigned to
High School Degree
Four-Year College Degree



The bar is composed of four horizontal segments: two yellow and two blue. The blue segments are stacked on top of each other, while the yellow segments are stacked below them.

Among the bottom 75%
of the high school class

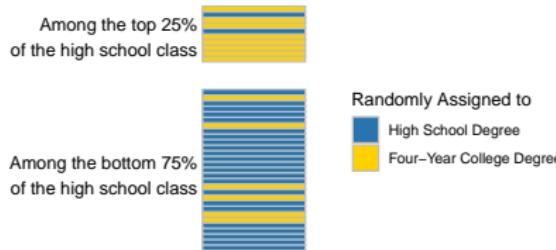
Conditional randomization: Exchangeability does not hold

Treated units are more likely to have done well in high school

Those who do well in high school are more likely to be employed at age 40 even without college

$$\{Y^1, Y^0\} \not\perp\!\!\!\perp A$$

A hypothetical experiment:
Conditional randomization



Conditional randomization: Analyze within subgroups

A hypothetical experiment:
Conditional randomization

Among the top 25%
of the high school class



Among the bottom 75%
of the high school class



Randomly Assigned to

- High School Degree
- Four-Year College Degree

Conditional randomization: Analyze within subgroups

Among top 25%, simple random experiment.

Among bottom 75%, simple random experiment.

A hypothetical experiment:
Conditional randomization

Among the top 25%
of the high school class



Among the bottom 75%
of the high school class



Randomly Assigned to

- High School Degree
- Four-Year College Degree

Conditional randomization: Analyze within subgroups

Among top 25%, simple random experiment.

Among bottom 75%, simple random experiment.

Conditional exchangeability:

$$\underbrace{\{Y^1, Y^0\}}_{\text{Potential Outcomes}} \perp\!\!\!\perp \underbrace{\text{Are Independent of}}_{A} \underbrace{\text{Treatment}}_{\text{Within Subgroups}} \mid \underbrace{X}_{\text{of } X}$$

A hypothetical experiment:
Conditional randomization



Randomly Assigned to

- High School Degree
- Four-Year College Degree

Among the bottom 75%
of the high school class



Conditional average treatment effects

We get two estimates. Average effect of college on employment

- ▶ among those in the top 25% of their high school class
- ▶ among those in the bottom 75% of their high school class

These are **conditional average treatment effects**

$$\underbrace{\tau(x)}_{\text{Conditional Average Treatment Effect (CATE)}} = \underbrace{E}_{\text{Expected value of}} \left(\underbrace{Y^1 - Y^0}_{\text{treatment effect}} \mid \underbrace{\vec{X} = \vec{x}}_{\text{the predictors } \vec{X} \text{ take the value } \vec{x}} \right)$$

Effect heterogeneity: CATEs differ across subgroups

Why might the effect of college on future employment

- ▶ be larger for those from the top 25% of the high school class?
- ▶ be larger for those from the bottom 75% of the high school class?

Effect heterogeneity and policy

Suppose we study (college → employment) in two subgroups

- ▶ Advantaged subgroup
 - ▶ Both parents finished college
 - ▶ Top quartile of family income at age 14
 - ▶ Took college prep courses
- ▶ Disadvantaged subgroup
 - ▶ Neither parent finished college
 - ▶ Bottom quartile of family income at age 14
 - ▶ Took college prep courses

Discuss:

1. Whose CATE would be larger?
2. How might the difference inform policy?

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