

What is Your Estimand?

A workshop on
specifying quantitative
research goals



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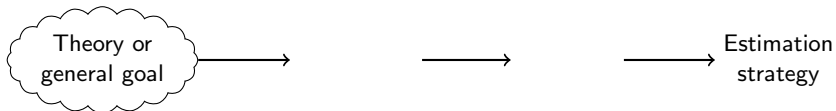
NOTE: These slides are for a pedagogical workshop.
The main slides for this paper are [here](#).

15 October 2021. Cornell Population Center Training Workshop.
Paper in *American Sociological Review*. Preprint on [SocArxiv](#). Replication code on [Dataverse](#). Research reported in this publication was supported by The Eunice Kennedy Shriver National Institute of Child Health & Human Development of the National Institutes of Health under Award Number P2CHD047879.

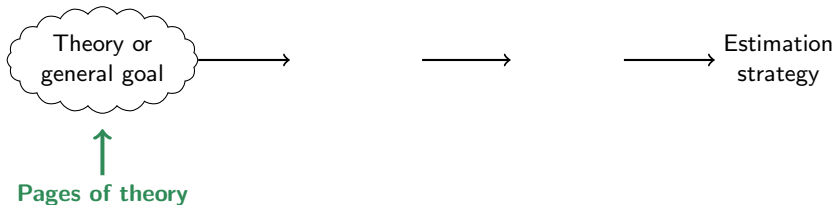
Plan for today

- ▶ Brief recap of paper: “What is Your Estimand?”
- ▶ In groups, we will discuss
 - ▶ Defining the intervention in a causal estimand
 - ▶ Defining the target population in a descriptive estimand
 - ▶ Policy implications of descriptive and causal estimands
- ▶ Professional development: Open for questions

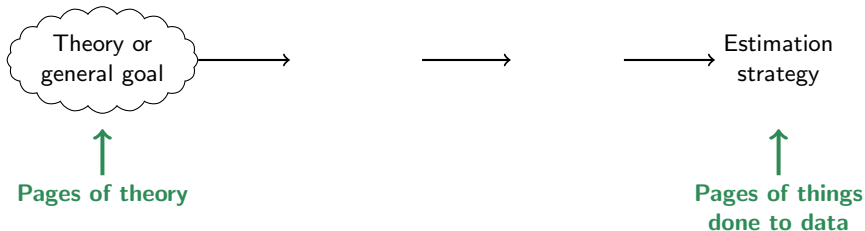
Research framework: Estimands connect theory to evidence



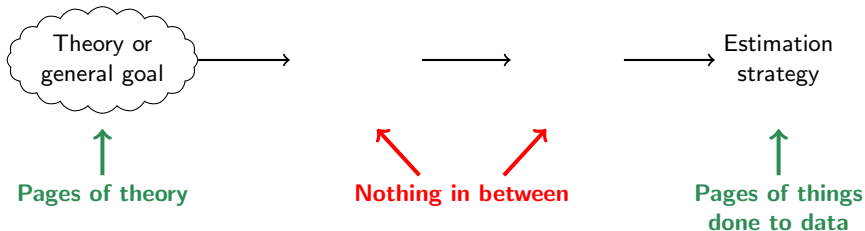
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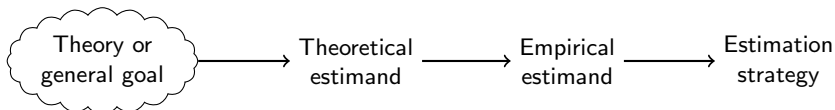
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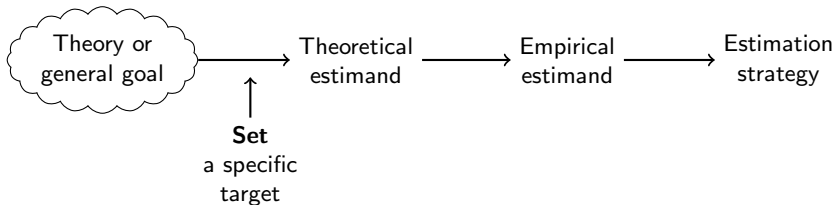
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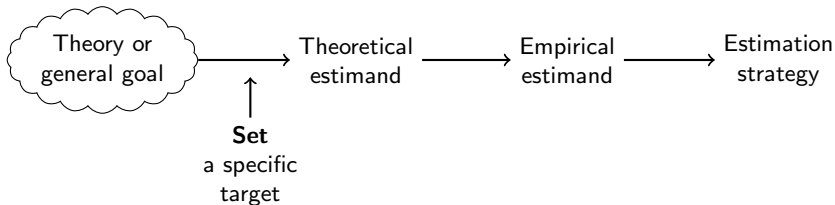
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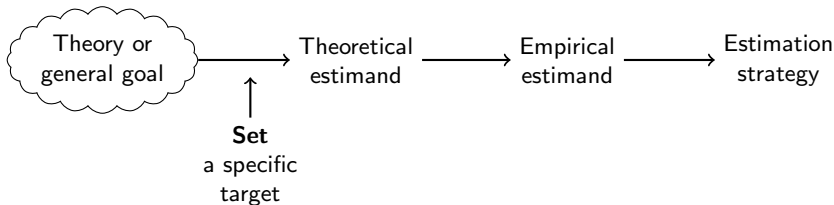
Research framework: Estimands connect theory to evidence



Definition

A **unit-specific quantity**
aggregated over a
target population

Research framework: Estimands connect theory to evidence



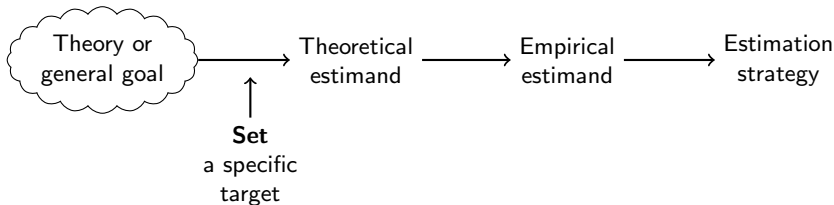
Definition

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Example

$$\frac{1}{\text{Size of U.S. adult population}} \sum_{i \text{ in U.S. adult population}} \left(\text{Employed}_i \right)$$

Research framework: Estimands connect theory to evidence



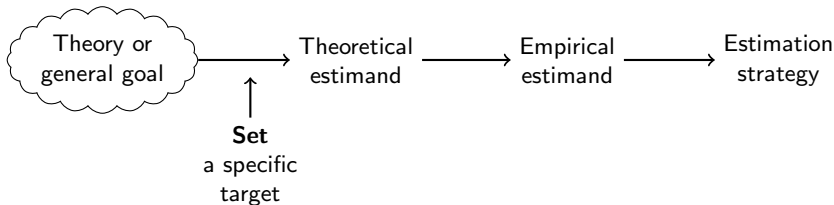
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$$\frac{1}{\text{Size of U.S. adult population}} \sum_{i \text{ in U.S. adult population}} \left(\underbrace{\text{Employed}_i(\text{Job training})}_{\text{Employment if received job training}} - \underbrace{\text{Employed}_i(\text{No job training})}_{\text{Employment if did not receive job training}} \right)$$

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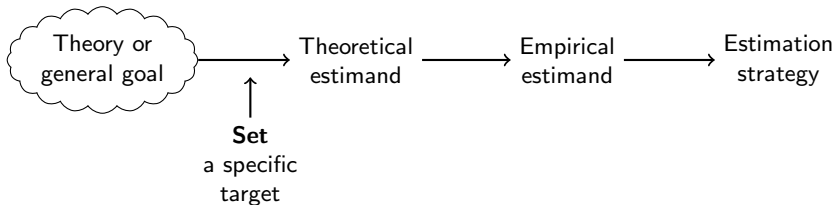
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Liebertson 1987, Abbott 1988, Freedman 1991, Xie 2013, Hernán 2018

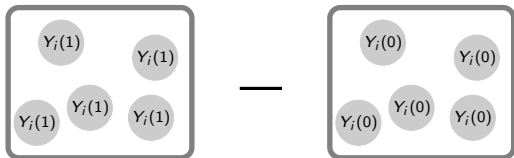
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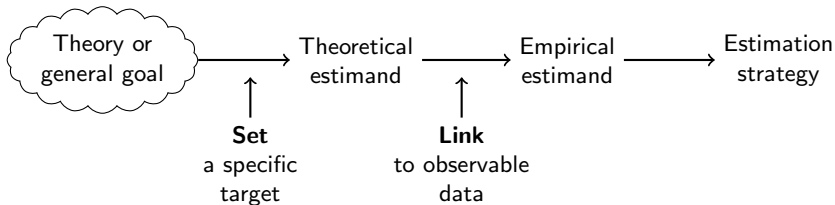
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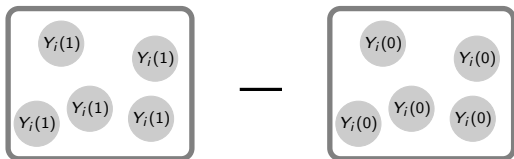
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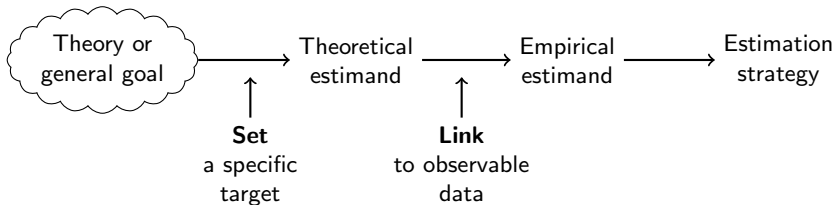
Definition

A quantity involving
observable data

Example



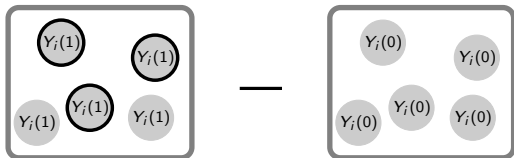
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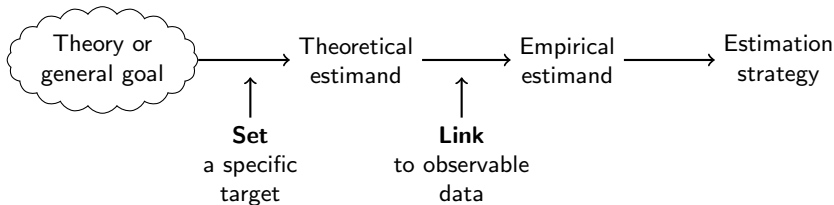
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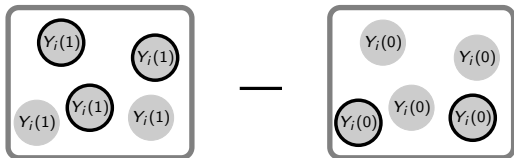
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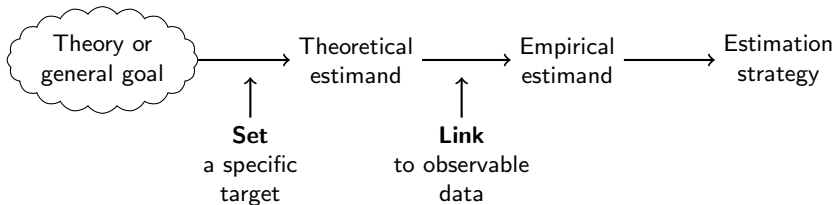
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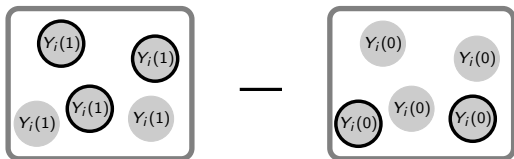
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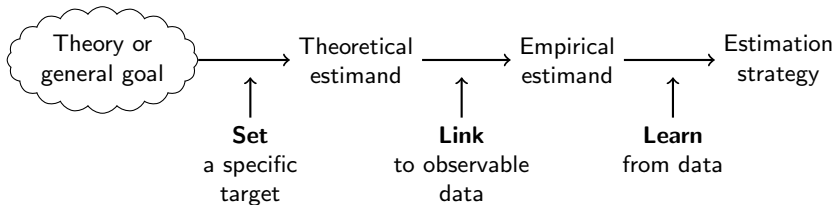


$$\vec{X} \xrightarrow{\quad} T \xrightarrow{\quad} Y$$

A curved arrow also points from \vec{X} to Y , representing a direct relationship or transformation.

Pearl 2009, Imbens and Rubin 2015,
Morgan and Winship 2015, Elwert and Winship 2014

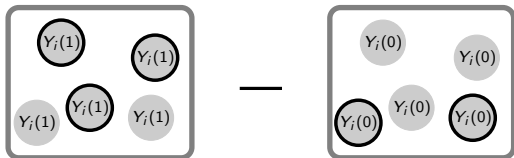
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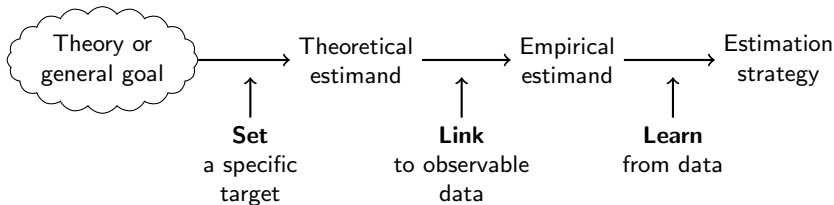
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An algorithm applied to data

Example



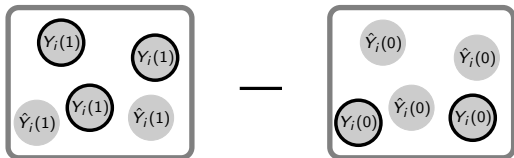
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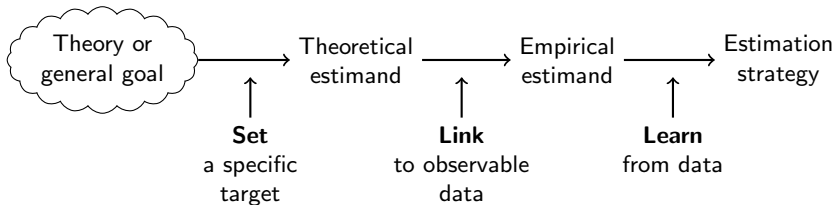
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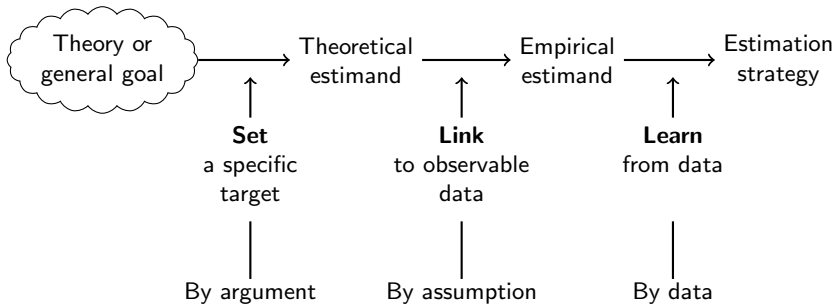
Example



Research framework: Estimands connect theory to evidence



Research framework: Estimands connect theory to evidence





1. Set the target quantity.



Describe a population

What is the proportion employed
among U.S. resident women ages 21–35?



Describe a population

What is the proportion employed
among U.S. resident women ages 21–35?

Woman 1

Woman 2

Woman 3

Woman 4



Describe a population

What is the proportion employed
among U.S. resident women ages 21–35?

	<u>Employed?</u>
Woman 1	1
Woman 2	0
Woman 3	1
Woman 4	1



Describe population subgroups

What is the proportion employed among U.S. resident women ages 21–35, comparing mothers to non-mothers?



Describe population subgroups

What is the proportion employed among U.S. resident women ages 21–35, comparing mothers to non-mothers?

	<u>Employed?</u>		<u>Employed?</u>
Mother 1	0	Non-Mother 1	1
Mother 2	0	Non-Mother 2	0
Mother 3	0	Non-Mother 3	1
Mother 4	1	Non-Mother 4	1



Causal effect in a population

What is the causal effect of motherhood on employment among U.S. resident women ages 21–35?



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Woman 1

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Causal effect in a population

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	Would be employed if a mother? $Y(1)$
Woman 1	0
Woman 2	0
Woman 3	0
Woman 4	1



Causal effect in a population

What is the causal effect of motherhood on employment among U.S. resident women ages 21–35?

	Would be employed if a mother? $Y(1)$	Would be employed if a non-mother? $Y(0)$
Woman 1	0	1
Woman 2	0	0
Woman 3	0	1
Woman 4	1	1



Causal effect in a population

What is the causal effect of motherhood on employment among U.S. resident women ages 21–35?

	Would be employed if a mother? $Y(1)$	Would be employed if a non-mother? $Y(0)$	Causal effect $Y(1) - Y(0)$
Woman 1	0	1	-1
Woman 2	0	0	0
Woman 3	0	1	-1
Woman 4	1	1	0



Describe population subgroups

What is the proportion employed among U.S. resident women ages 21–35, comparing mothers to non-mothers?

	Employed?		Employed?
Mother 1	0	Non-Mother 1	1
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Causal effect in a population

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Woman 2	0	0	0
Woman 3	0	1	-1
Woman 4	1	1	0

Very
different
research
goals



If the estimand is causal, you have to tell the reader:

what is the intervention?



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Example: Effect of motherhood on employment



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Example: Effect of motherhood on employment

— How many kids?



If the estimand is causal, you have to tell the reader:

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Example: Effect of motherhood on employment

- How many kids?
- Adopted? Biological?



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Example: Effect of motherhood on employment

- How many kids?
- Adopted? Biological?
- Sex mix of the kids?



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Example: Effect of motherhood on employment

- How many kids?
- Adopted? Biological?
- Sex mix of the kids?
- Born this year? 5 years ago?



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Example: Effect of motherhood on employment

- How many kids?
- Adopted? Biological?
- Sex mix of the kids?
- Born this year? 5 years ago?

One resolution: Effect **among mothers** with their factual configuration of kids, compared with a counterfactual of having no kids



Group Exercise

If the estimand is causal, you have to tell the reader:

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Group Exercise

If the estimand is causal, you have to tell the reader:

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“If poor women would marry,
they would no longer be poor.”

— A policymaker



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$$\frac{1}{n} \sum_{i=1}^n \left(\text{Poverty}_i(\text{Married}) - \text{Poverty}_i(\text{Unmarried}) \right)$$



Group Exercise

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Is this policymaker being sufficiently precise about the intervention?



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Is this policymaker being sufficiently precise about the intervention?

Suggestion: For a given woman i , are there versions of “married” and “unmarried” that might lead to distinct outcomes?
Does it matter who you marry?



Regardless of descriptive or causal,
you have to tell the reader the **target population**



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Example: The unemployment rate



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Example: The unemployment rate

U.S. non-institutionalized civilians age 16+
actively looking for work in past 4 weeks



Group Exercise

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Group Exercise

Regardless of descriptive or causal,
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“My online sample is
nationally representative”

— An academic



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What might the target population be if the question is:



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What might the target population be if the question is:

1) Do you plan to vote for Biden?



Group Exercise

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What might the target population be if the question is:

- 1) Do you plan to vote for Biden?
- 2) What is your hourly wage?



Group Exercise

Regardless of descriptive or causal,
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— An academic

What might the target population be if the question is:

- 1) Do you plan to vote for Biden?
- 2) What is your hourly wage?
- 3) Would you recommend this hypothetical resume for an interview?



Policy implications depend on the type of claim.



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Descriptive Claim

“People who go to Disneyland twice are more likely to buy an annual pass”



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Policy implications for Disney



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Descriptive Claim

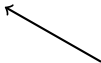
"People who go to Disneyland twice are more likely to buy an annual pass"

Target annual pass ads at the return visitors

Causal Claim

"A second trip to Disneyland causes people to decide they need an annual pass"

Policy implications for Disney





Policy implications depend on the type of claim.

Descriptive Claim

"People who go to Disneyland twice are more likely to buy an annual pass"

Target annual pass ads at the return visitors

Causal Claim

"A second trip to Disneyland causes people to decide they need an annual pass"

Offer discounts to induce a second visit

Policy implications for Disney





Group Exercise

Policy implications depend on the type of claim.

Descriptive Claim

Causal Claim



Group Exercise

Policy implications depend on the type of claim.

Descriptive Claim

“College graduates are more financially stable than non-graduates”

Causal Claim



Group Exercise

Policy implications depend on the type of claim.

Descriptive Claim

"College graduates are more financially stable than non-graduates"

Causal Claim

"College completion causes greater financial stability"



Group Exercise

Policy implications depend on the type of claim.

Descriptive Claim

"College graduates are more financially stable than non-graduates"

Causal Claim

"College completion causes greater financial stability"

What kinds of policies would these claims suggest?

Professional development

- ▶ Social media
- ▶ Professional networking
- ▶ Anything else you want to discuss!