Machine learning for population description

lan Lundberg UCLA Including past work with Rebecca Johnson (Georgetown) Brandon Stewart (Princeton)

and current work with Kristin Liao (UCLA) ilundberg.github.io/description

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Plan for today

- ► Estimands in quantitative social science
- ightharpoonup Descriptive estimands: A \hat{Y} view
- ► Intro to tomorrow's computer tutorial

What Is Your Estimand? **Defining the Target Quantity** Connects Statistical Evidence to Theory

American Sociological Review

1 - 34

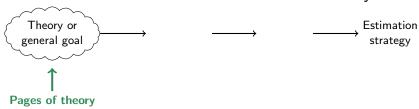
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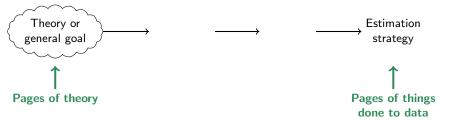
DOI:10.1177/00031224211004187 journals.sagepub.com/home/asr

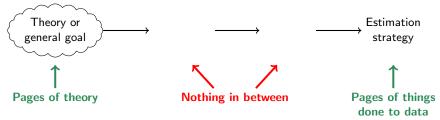


Ian Lundberg, a Rebecca Johnson, b and Brandon M. Stewarta

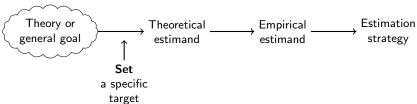
Theory or general goal Strategy

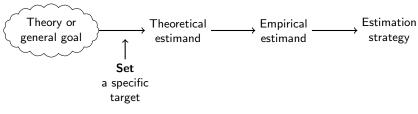






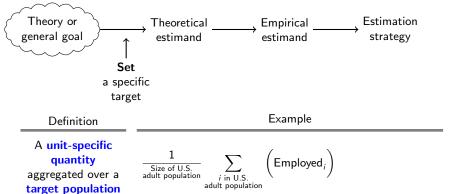
 $\begin{array}{c} \text{Theory or} \\ \text{general goal} \end{array} \xrightarrow{\text{Theoretical}} \xrightarrow{\text{estimand}} \xrightarrow{\text{estimand}} \xrightarrow{\text{Estimation}}$

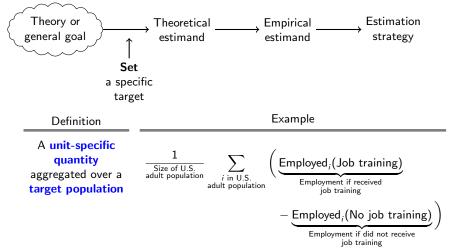


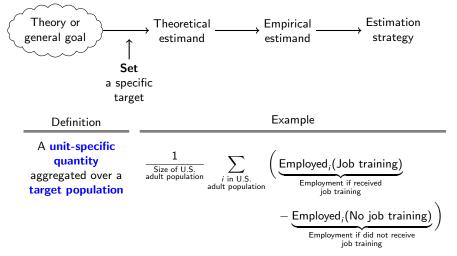


Definition

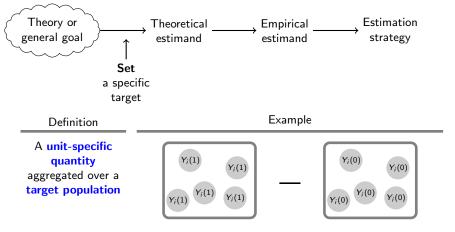
A unit-specific quantity aggregated over a target population

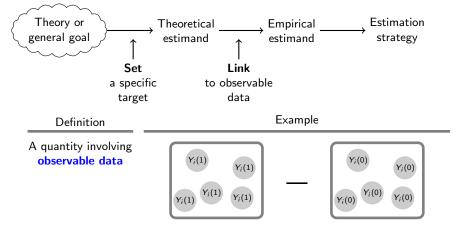


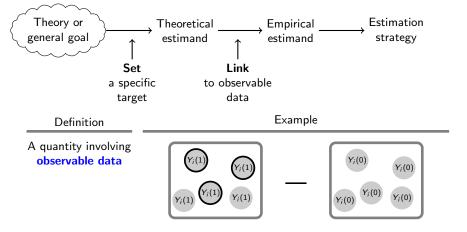


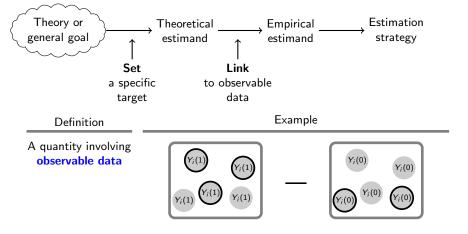


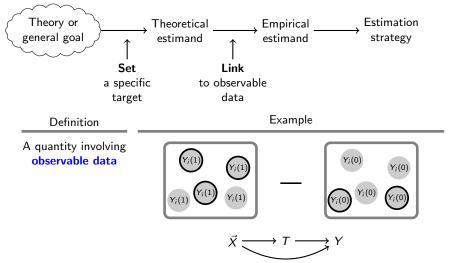
Lieberson 1987, Abbott 1988, Freedman 1991, Xie 2013, Hernán 2018



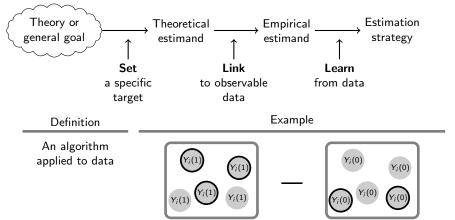


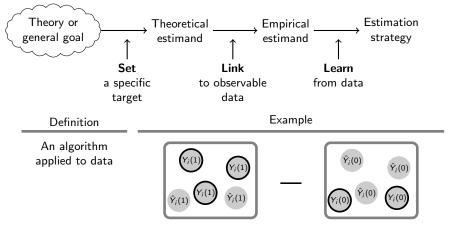


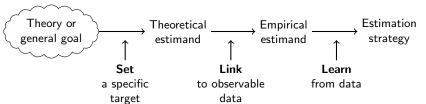


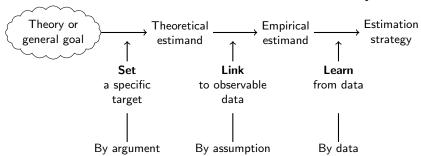


Pearl 2009, Imbens and Rubin 2015, Morgan and Winship 2015, Elwert and Winship 2014









	Theory or	Set	Theoretical	Link	Empirical	Learn	Estimation
,	general goal	by argument	estimand	by assumption	estimand	by data	strategy

Theory or general goal by argument Theoretical estimand by assumption Estimation by data Estimation

Effect of motherhood on employment

Theory or general goal by argument Theoretical estimand by assumption Estimation by data Estimation

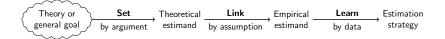
Effect of motherhood on employment

First two births are the same sex



Effect of motherhood on employment

First two births are the same sex Third birth



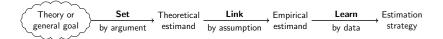
Effect of motherhood on employment

First two births are the same sex
$$\longrightarrow$$
 Third birth \longrightarrow Employed



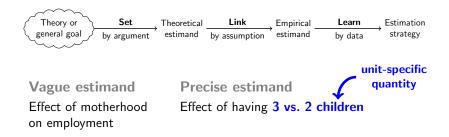
Vague estimand Effect of motherhood on employment

First two births are the same sex
$$\longrightarrow$$
 Third birth \longrightarrow Employed

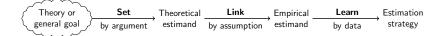


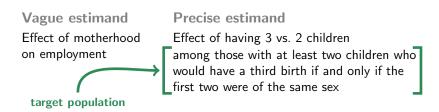
Vague estimand Effect of motherhood on employment Precise estimand

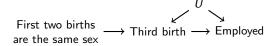
First two births are the same sex
$$\longrightarrow$$
 Third birth \longrightarrow Employed



First two births are the same sex \longrightarrow Third birth \longrightarrow Employed







Theory or general goal by argument Theoretical Link by assumption estimand by data Estimation by data

Precise estimand

Effect of having 3 vs. 2 children among those with at least two children who would have a third birth if and only if the first two were of the same sex

 \approx 4% of all mothers

Theory or general goal by argument estimand by assumption estimated by data Estimation strategy

Precise estimand

Effect of having 3 vs. 2 children among those with at least two children who would have a third birth if and only if the first two were of the same sex

 \approx 4% of all mothers

You have to argue either:

- 1)
- 2)

Theory or general goal by argument estimand by assumption estimand by data Estimation

Precise estimand

Effect of having 3 vs. 2 children among those with at least two children who would have a third birth if and only if the first two were of the same sex

 \approx 4% of all mothers

You have to argue either:

- 1) That estimand matters for theory, or
- 2)

Theory or general goal by argument estimand by assumption estimand by data Estimation

Precise estimand

Effect of having 3 vs. 2 children among those with at least two children who would have a third birth if and only if the first two were of the same sex

$$\approx$$
 4% of all mothers

You have to argue either:

- 1) That estimand matters for theory, or
- 2) It speaks to some broader estimand



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?

_	Employed?
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?

Theory or	Set	Theoretical	Link	Empirical	Learn	Estimation
general goal	by argument	estimand	by assumption	estimand	by data	strategy

Describe population subgroups

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

Employed?		Employed?
0	Non-Mother 1	1
0	Non-Mother 2	0
0	Non-Mother 3	1
1	Non-Mother 4	1
	0 0 0 0 1	0 Non-Mother 1 0 Non-Mother 2 0 Non-Mother 3





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

Woman 1

Woman 2

Woman 3

Woman 4

Theory or	Set	Theoretical	Link	Empirical	Learn	Estimation
general goal	by argument	estimand	by assumption	estimand	by data	strategy

Mould be

	vvouid be	
	employed if	
	a mother?	
	Y(1)	
Woman 1	0	
vvoman 1	U	
Woman 2	0	
Woman 3	0	
Woman 4	1	

Theory or	Set	Theoretical	Link	Empirical	Learn	Estimation
general goal	by argument	estimand	by assumption	estimand	by data	strategy

	Would be	Would be
	employed if	employed if
	a mother?	a non-mother
	Y(1)	<i>Y</i> (0)
Woman 1	0	1
Woman 2	0	0
Woman 3	0	1
Woman 4	1	1

Theory or	Set	Theoretical	Link	Empirical	Learn	Estimation
general goal	by argument	estimand	by assumption	estimand	by data	strategy

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

Why model?

Causal effect in a population

	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

Why model?

Causal effect in a population

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

Why model?

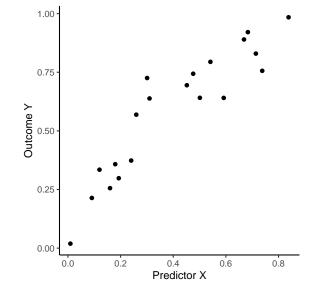
Describe population subgroups

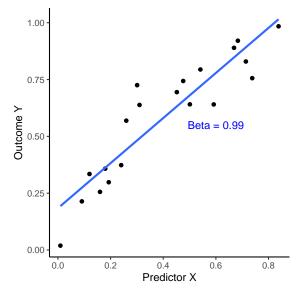
What is the proportion employed among U.S. resident women ages 21–35, comparing mothers to non-mothers?

-	Employed?	_	Employed?
Mother 1	0	Non-Mother 1	1
Mother 2	0	Non-Mother 2	0
Mother 3	0	Non-Mother 3	1
Mother 4	1	Non-Mother 4	1

A \hat{Y} view of description With Kristin Liao, UCLA



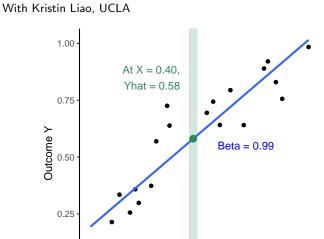




A \hat{Y} view of description

0.00

0.0



0.2

0.4

Predictor X

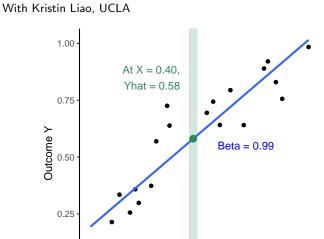
0.6

0.8

A \hat{Y} view of description

0.00

0.0



0.2

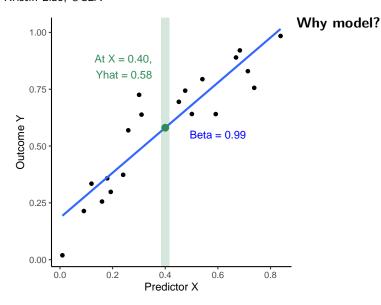
0.4

Predictor X

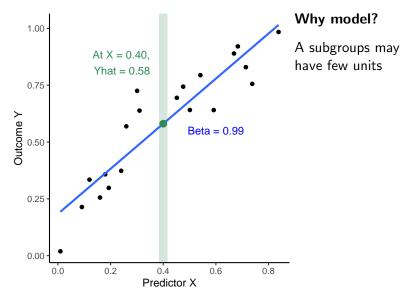
0.6

0.8

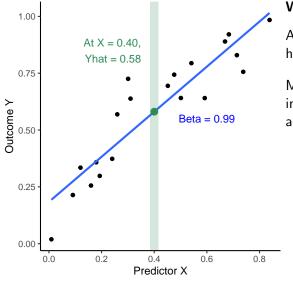
A \hat{Y} view of description With Kristin Liao, UCLA



A \hat{Y} view of description



With Kristin Liao, UCLA

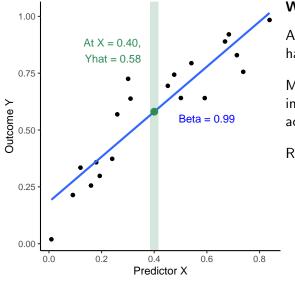


Why model?

A subgroups may have few units

Model pools information across subgroups

With Kristin Liao, UCLA

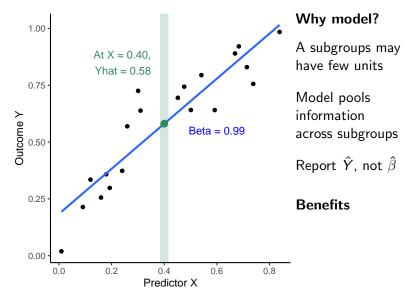


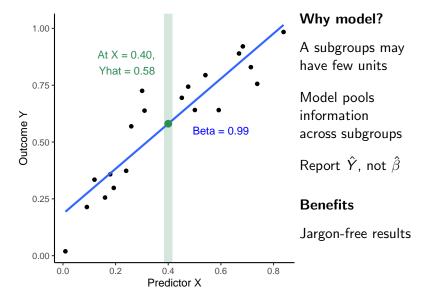
Why model?

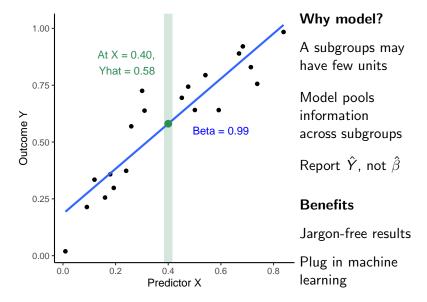
A subgroups may have few units

Model pools information across subgroups

Report \hat{Y} , not $\hat{\beta}$







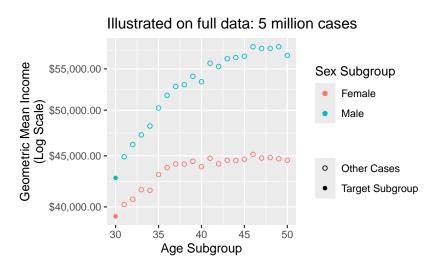
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Sample of 5 million cases (true nonparametric estimates)
Simulate a sample of 100 (evaluate sample-based estimators)

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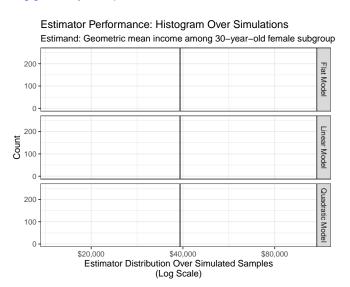
Source of 5 million cases

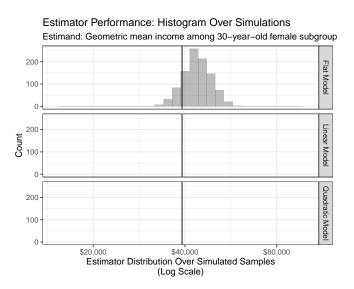
- ► American Community Survey (ACS) 2010–2019
- ► Adults age 30–50
- ► Worked 35+ hours per week in 50+ weeks last year
- ► Outcome: Annual wage and salary income

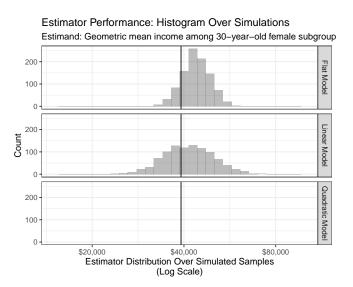


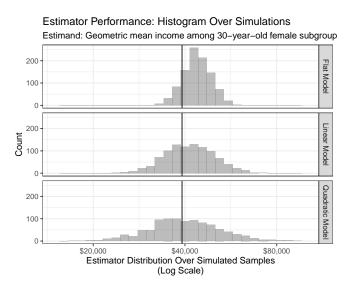


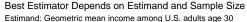


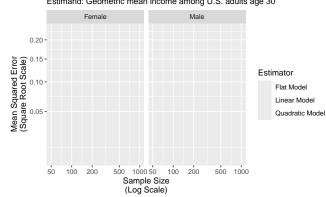


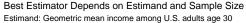


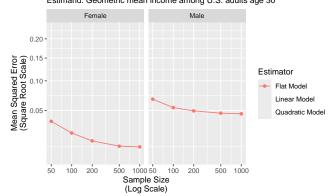




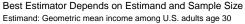


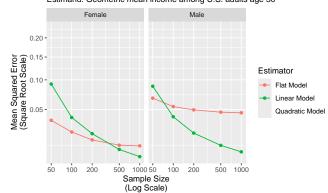




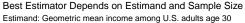


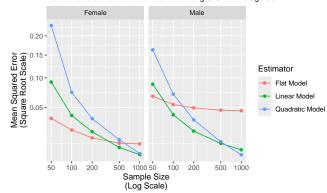
Evaluate models



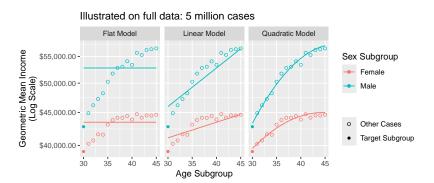


Evaluate models





Evaluate models



- ▶ a model as a means to an end
 - ▶ we would rather not model
 - ► model only when you lack data

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- ▶ a model as a means to an end
 - ▶ we would rather not model
 - ► model only when you lack data
- ► misspecified models are ok
 - ► flat model was wrong
 - ► flat model was best

(lower variance)

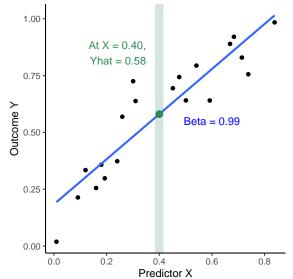
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- ▶ a model as a means to an end
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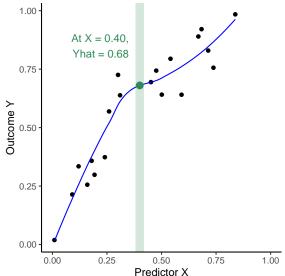
(lower variance)

► machine learning becomes a plug-in

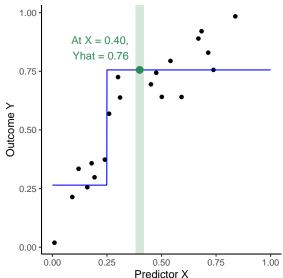
With \hat{Y} description, machine learning becomes a plug-in



With \hat{Y} description, machine learning becomes a plug-in



With \hat{Y} description, machine learning becomes a plug-in



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We will give you data:

▶ male and female incomes at age 30–50 in 2010–2019

You will make a forecast:

▶ male and female geometric mean income at age 30–50 in 2022

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Prepare the environment by loading the tidyverse package.

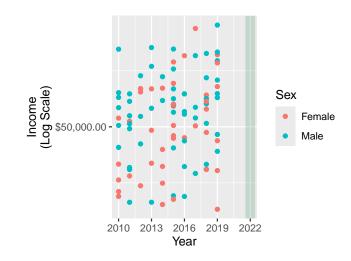
```
library(tidyverse)
```

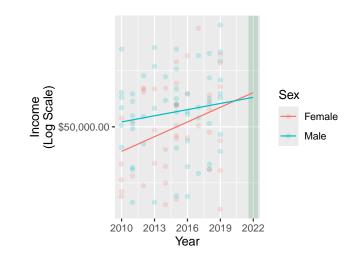
The function below simulates a sample of 100 cases.

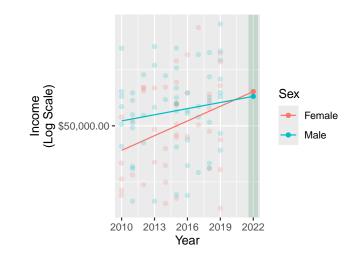
```
simulate <- function(n = 100) {
  read_csv("https://ilundberg.github.io/description/assets/truth.csv") |>
    slice_sample(n = n, weight_by = weight, replace = T) |>
    mutate(income = exp(rnorm(n(), meanlog, sdlog))) |>
    select(year, age, sex, income)
}
```

We can see how it works below.

```
simulated <- simulate(n = 100)</pre>
```







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We will give you data:

▶ male and female incomes at age 30–50 in 2010–2019

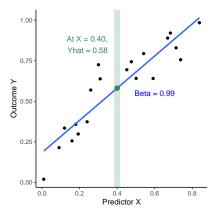
You will make a forecast:

▶ male and female geometric mean income at age 30–50 in 2022

We will see who comes closest

► to gold-standard truth from ACS 2022

Thanks!



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