

# FY22 Budget Training

*using and generating evidence*



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Rhode Island

Director, The Policy Lab  
[thepolicylab.brown.edu](https://thepolicylab.brown.edu)

## learning objectives

- How do you assess the existing evidence base?
  - Methods 101 crash course
  - The Evidence Scale
- When and how might you invest in generating new evidence?
  - Recall other resources:
    - Budget Instructions
    - Decision Package Template
    - Examples (good and bad)
    - Website Trainer
    - Office Hours

what evidence do you need?

### **3. Opportunity Statement**

*In this section, clearly explain the problem that exists today and the opportunity that your request presents to capitalize on. The best opportunity statements thoroughly explain, with as much detail as possible: (1) where we are today; (2) where we want to be in the future; and (3) why there is the gap between where we are and where we want to be. The best opportunity statements also quantify key variables wherever possible.*

### **4. Proposed Intervention & Theory of Change**

*Provide a detailed description of the initiative you are proposing to respond to the above-described problem/capitalize on the opportunity. Your narrative here should clearly describe how your intervention, if funded, could close the gap described above and achieve the desired future state.*

# types of metrics

<b>inputs</b>	<b>outputs</b>	<b>outcomes</b>	<b>impacts</b>
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# types of metrics

<b>inputs</b>	<b>outputs</b>	<b>outcomes</b>	<b>impacts</b>
staff facilities materials etc.	# people served # job apps submitted	# employed average wages	Increase in employment or wages <b>CAUSED</b> <b>BY</b> the program

# aims of evaluation

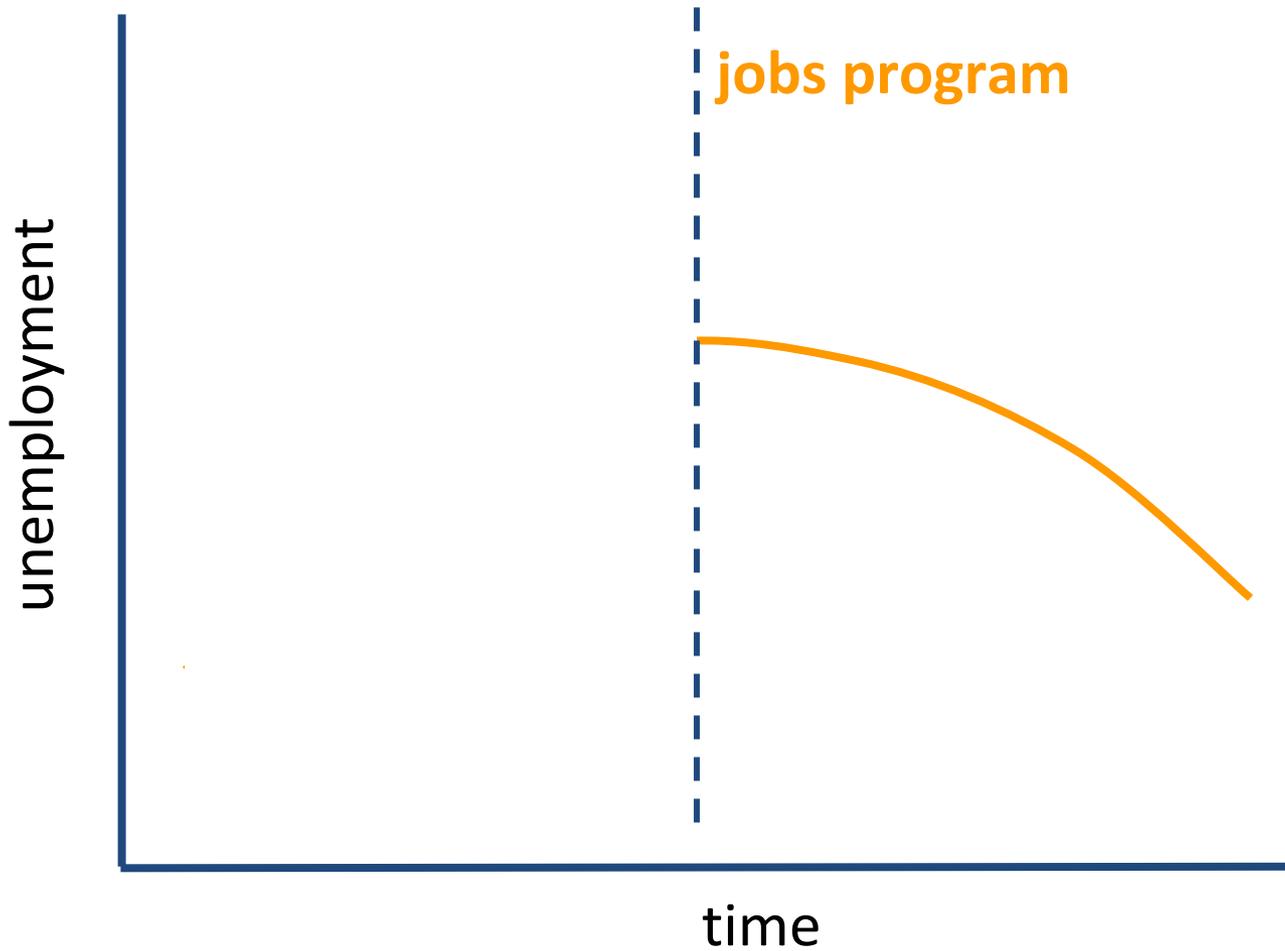
<b>inputs</b>	<b>outputs</b>	<b>outcomes</b>	<b>impacts</b>
staff facilities materials etc.	# people served # job apps submitted	# employed average wages	Increase in employment or wages <b>CAUSED BY</b> the program

**implementation**

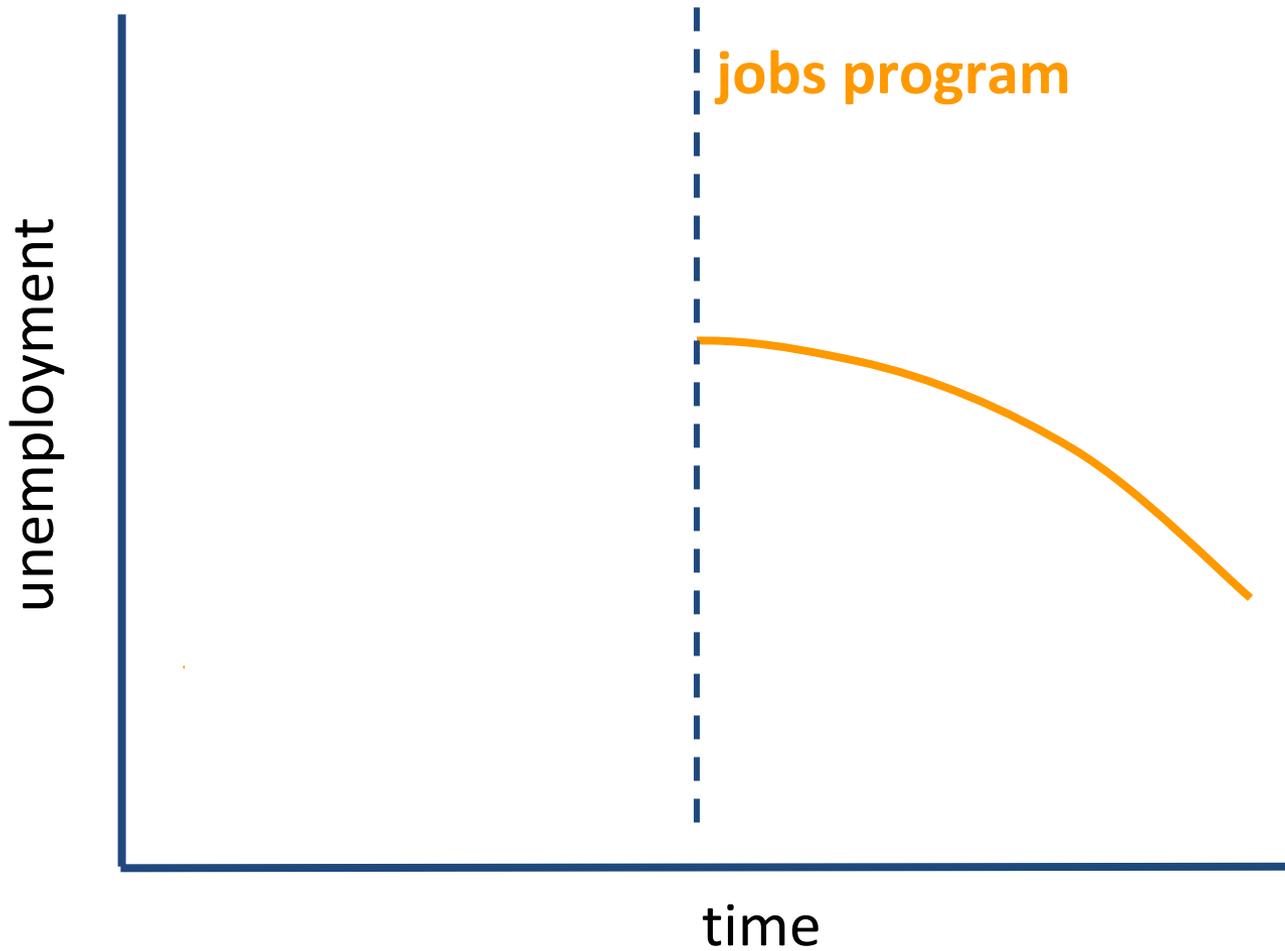
**outcome**

**impact**

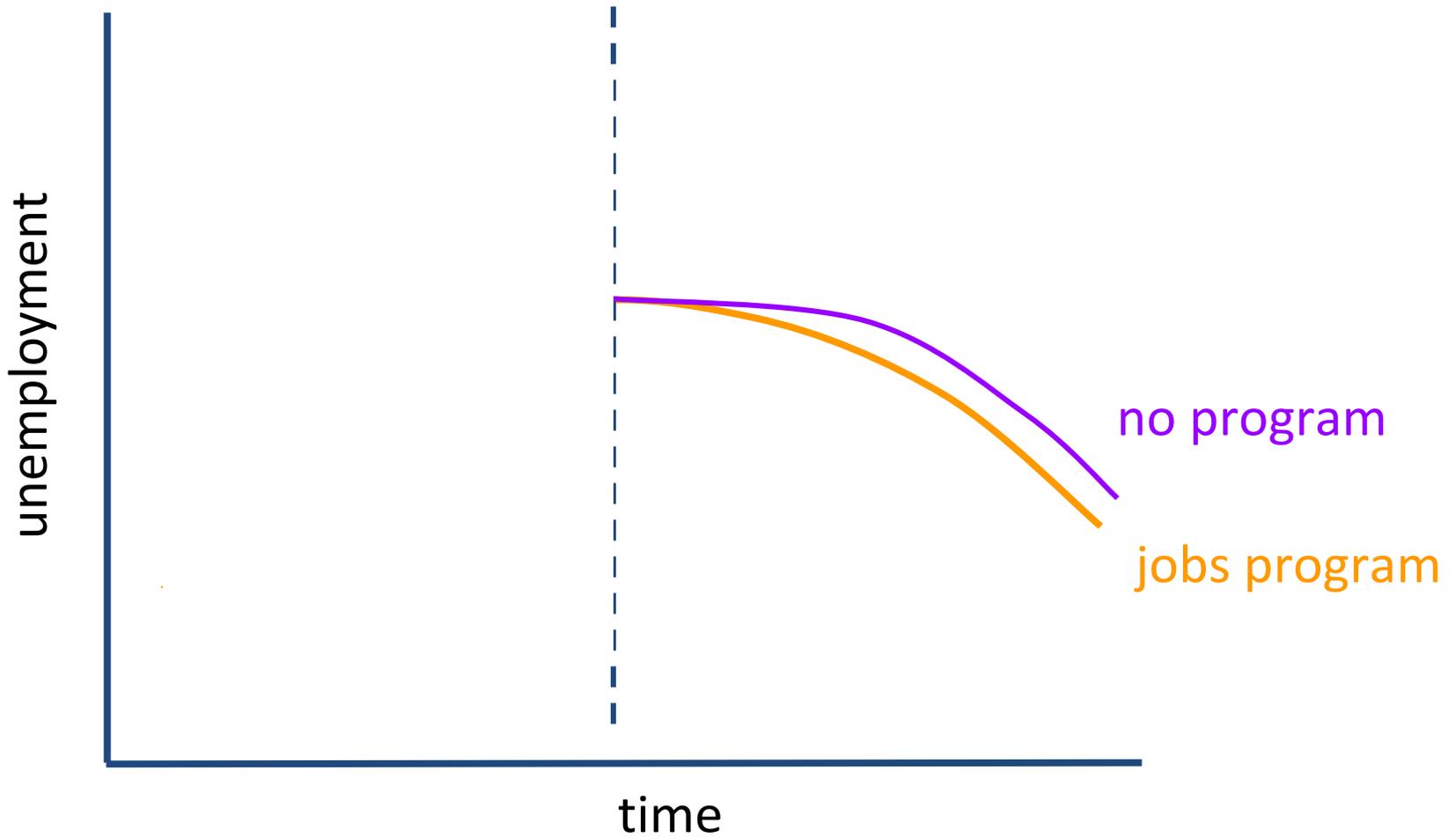
# impact / causal claims



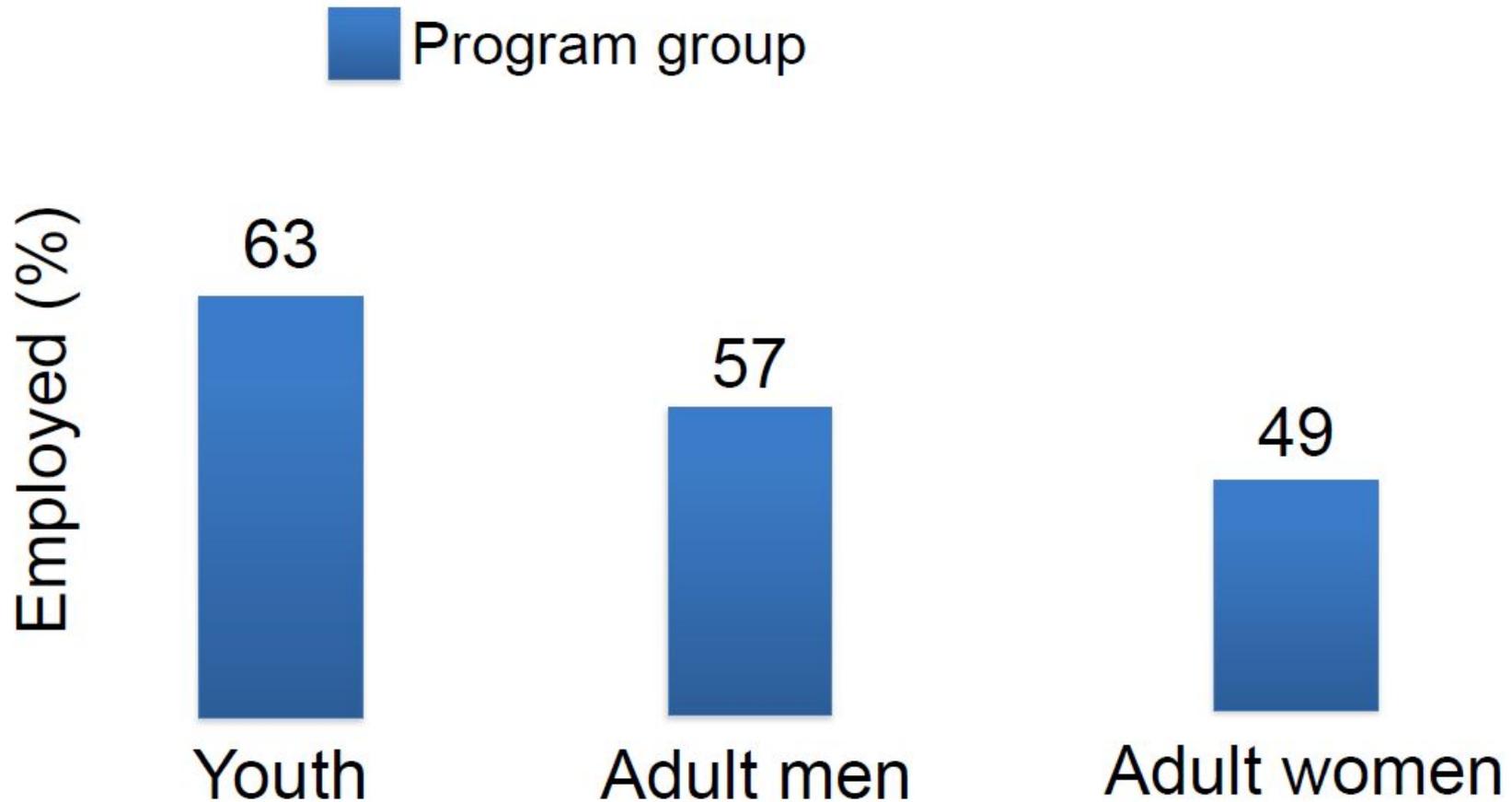
# CAUSAL CLAIMS



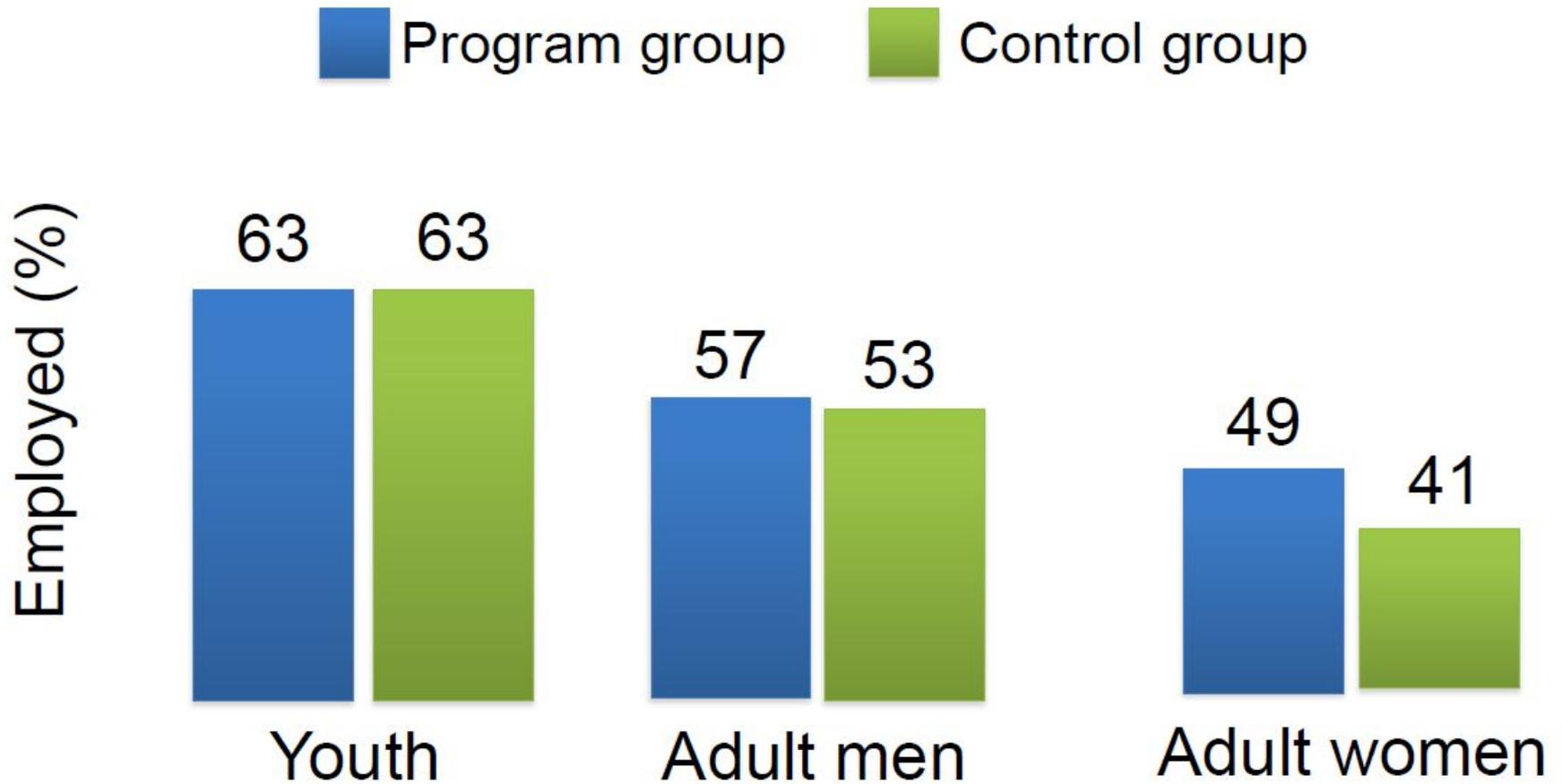
need a counterfactual



# who benefited the most from the jobs program?



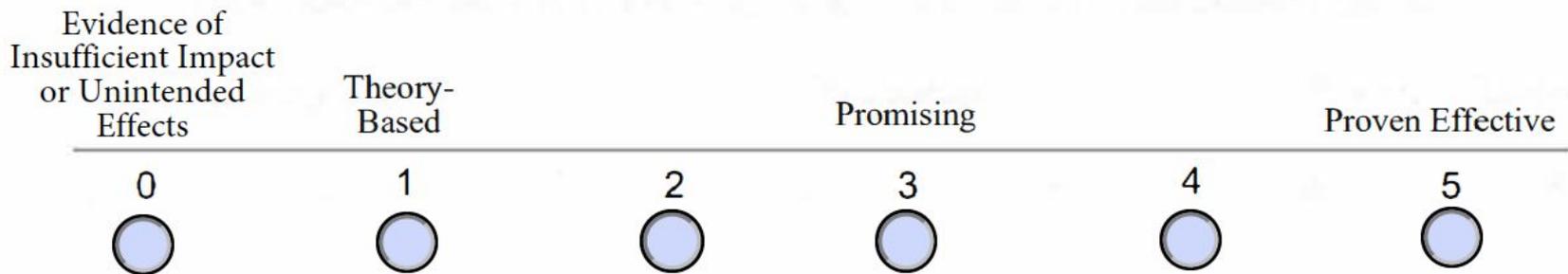
# who benefited the most from the jobs program?



finding the evidence you need

## 6. Evidence Scale Ranking

*Please rank the proposed initiative's current level of evidentiary support on a scale from 0-5, based on the [RI Evidence Scale](#), with one being the least evidentiary support and five being the most evidentiary support. You can use tools like the [Pew Results First Clearinghouse](#) and the [Social Programs That Work](#) database to determine whether the initiative you are proposing has been rigorously evaluated in other jurisdictions. The Office of Management & Budget understands that the majority of agency requests will likely not be in the top evidence tiers at the point of submittal, and you should certainly feel free to submit requests that are "theory-based" rather than evidence based. Please note that "theory-based" submissions should include a robust and compelling measurement and evaluation plan in the Performance Measurement section.*

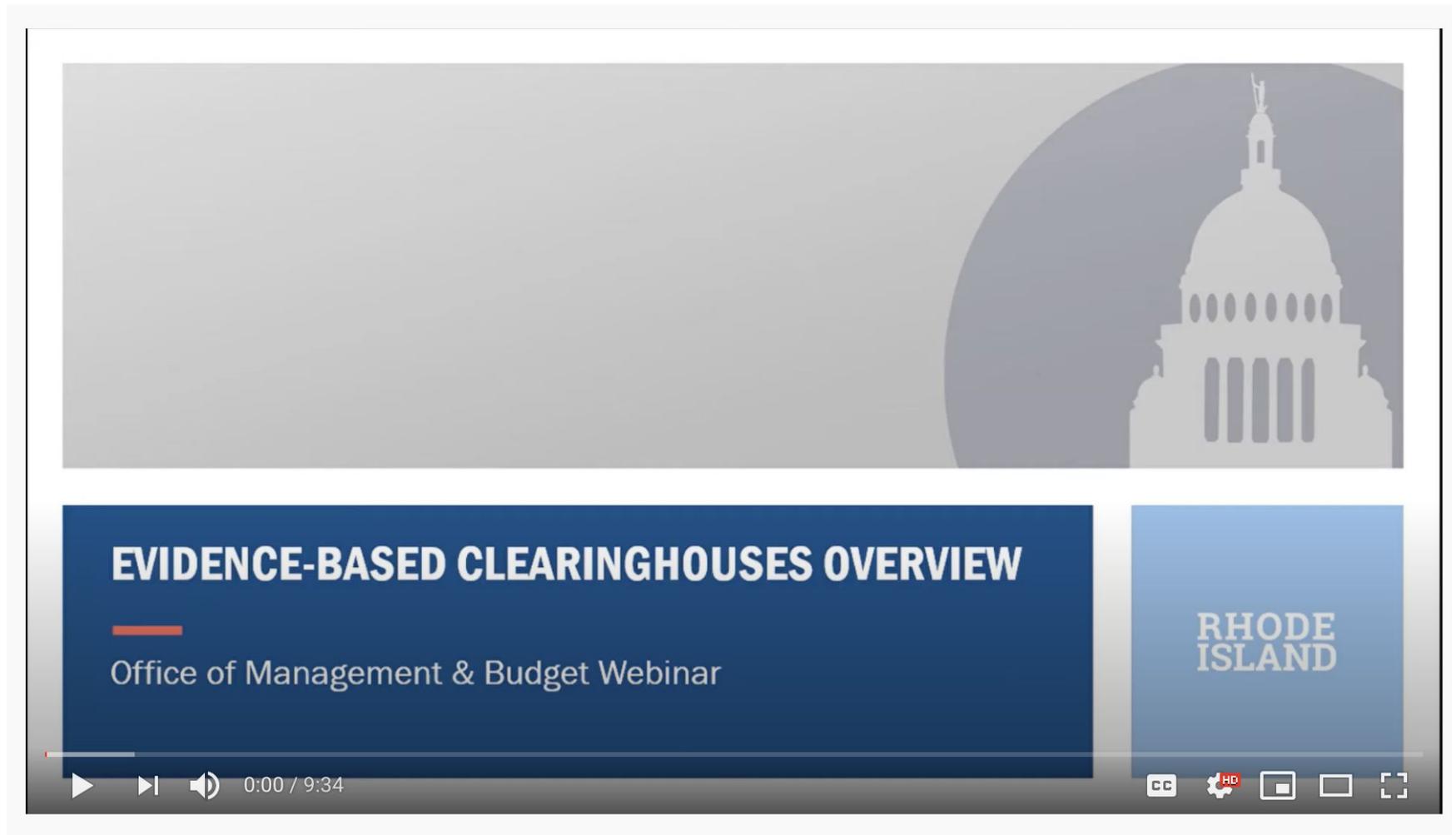


## 7. Description of Evidence Base

*Describe the justification for your evidence scale ranking. What evidence exists that makes you think that the proposed initiative will work? Where is there uncertainty of effectiveness? It is helpful to include citations, links, or attachments of relevant evidence source(s)*



# evidence clearinghouses



The image shows a YouTube video player interface. At the top, there is a large, faded image of a domed building, likely a state capitol. Below this, the video title "EVIDENCE-BASED CLEARINGHOUSES OVERVIEW" is displayed in white text on a dark blue background. Underneath the title, it says "Office of Management & Budget Webinar". To the right of the title, there is a blue square with the text "RHODE ISLAND" in white. At the bottom of the player, there is a progress bar showing "0:00 / 9:34" and several control icons including play, volume, and settings.

# 11 Tips for Doing Desk Research

## **WARNING: study quality**

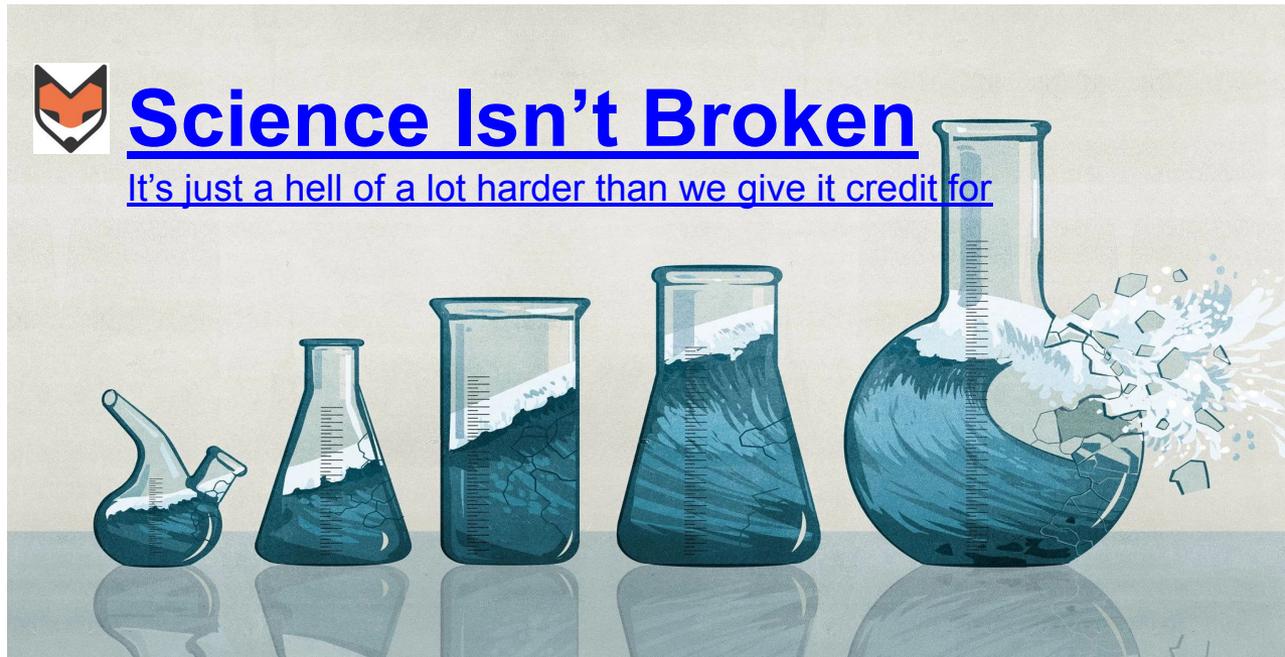
- Too few people were studied
- The people studied are not representative of the population you care about
- Causal claims made, but no counterfactual
- Effect sizes are not talked about clearly
- Fishy handling of the data

## **WARNING: study quality**

- Too few people were studied
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- **Fishy handling of the data**

# fishy handling of data

- File drawer problem
- $P$ -hacking



# pre-analysis plans help avoid fishiness

The screenshot shows the Open Science Framework (OSF) interface for a project titled "DC Body-Worn Camera Evaluation". The page is organized into several sections:

- Header:** "Open Science Framework" logo and navigation links: "Dashboard", "My Projects", "Browse", and a search icon. The user "David Yokum" is logged in.
- Project Title:** "DC Body-Worn Camera Evaluation" with a logo for "THE lab @ DC".
- Contributors:** David Yokum, Anita Ravishankar, Alexander Coppock, Heidi Fieselmann.
- Affiliated Institutions:** The Lab @ DC.
- Date created:** 2016-07-01 10:25 AM | Last Updated: 2017-02-01 01:36 PM.
- Category:** Project.
- Description:** A randomized controlled trial of the Police Body-Worn Camera Program of the Metropolitan Police Department of the District of Columbia.
- License:** No license.
- Wiki:** A section with the text "No wiki content".
- Files:** A section with a table of files. The table has columns for "Name" and "Modified". The files listed are "DC Body-Worn Camera Evaluation", "OSF Storage", and "Pre-Analysis Plan (PAP)".
- Citation:** A section with the URL "osf.io/p6vuh".
- Components:** A section with the text "No components have been added to this project." and buttons for "Add Component" and "Link Projects".
- Tags:** A section with a text input field containing "add a tag".

See <https://osf.io/yjyng/> for the publically pre-registered analysis plan. See also <https://osf.io/q6c45/> for PVD Talks example.

methods crash course

9. **What methodologies do you currently use for program evaluation? Check all that apply.**

- No evaluation is done
- Measurement of the resources (e.g. staff, material expenses) required to deploy the service or initiative
- Measurement of how many people use the service or initiative
- Focus groups, surveys, or other qualitative methods that ask people about their experience with the service/initiative
- Measurement of outcomes at a point in time or over time (e.g. monthly or quarterly reports of student test scores, average wages, crime reports, park admissions, medical claims, etc.)
- Measurement of causal impact with econometric methods (e.g. regression discontinuity, multiple regression, matched controls, instrumental variables)
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- Other:

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# outputs

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# impacts

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Other:

# wait, can't folk just self-report impact?

1. People may answer in ways they think you want, rather than what *they* really think.
2. People can't report unconscious causes of behavior.
3. Memory is faulty.
4. Often not a sufficiently representative sample.

## Industrial Funding Fee Reports

A confirmation prompt reduces financial self-reporting error

**Agency Objective.** Reduce financial self-reporting errors using a redesigned data-entry form.

**Background.** Federal vendors are required to pay a fee, called the industrial funding fee (IFF), currently set at 0.75 percent of quarterly sales on certain transactions.<sup>1</sup> The cost of the IFF generally is determined from self-reports submitted via a website (https://22.gsa.gov). In fiscal year 2013, the General Services Administration (GSA) collected approximately \$208 million in IFF across roughly 47,000 transactions.

The GSA introduced a confirmation prompt at the end of the IFF data-entry form in order to reduce financial self-reporting errors. Research has shown that viewing a confirmation prompt, where the user signs his or her name confirming the accuracy of the self-reported statements, reduces self-report errors if done at the beginning of a form; prompts at the end of a form seem to have no effect.<sup>2</sup>

**Methods.** The randomized controlled trial was fielded during the third reporting quarter of 2014, where vendors in 18,471 were randomly assigned to use either (A) the existing reporting system (control) or (B) a modified interface (treatment), redesigned to include an opening signature box confirming, "I promise that the information I am providing is true and accurate." Administrative data on paid IFF provided the primary outcome measure.

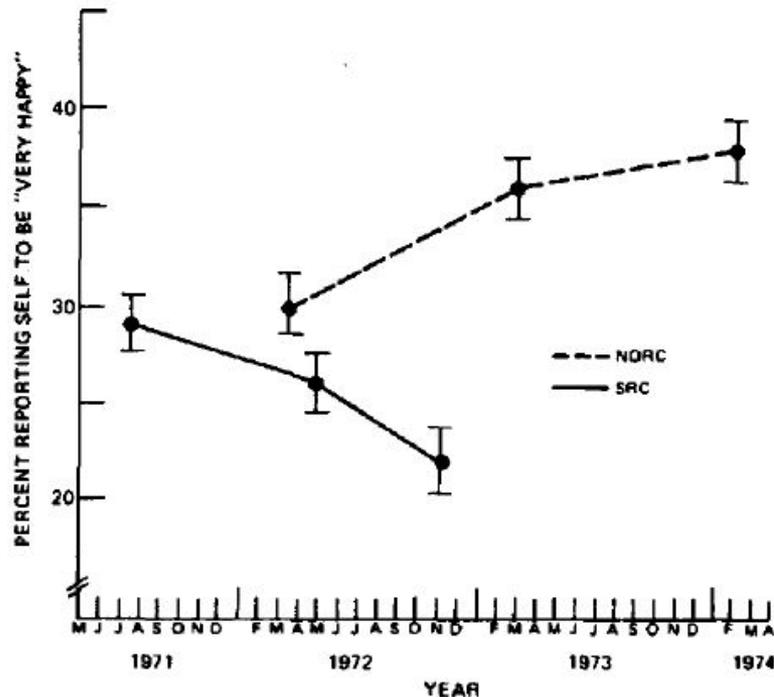
**Results.** The median self-reported sales amount was \$485 (μ = 0.05, 95% CI \$97, 803) higher for vendors signing at the top of the form compared with those vendors who were not required to make this confirmation. This increase in IFF remittance in the treatment group is just the third quarter of 2014 and \$1.59 million.

**Conclusions.** Confirmation prompts at the beginning of a form are a promising approach to reducing financial self-reporting errors, especially given the near-zero marginal cost to implement.

Quarter	Year	Amount	Rate	Payment
Q3	2014	\$1,000,000	0.0075	\$7,500
Q4	2014	\$1,000,000	0.0075	\$7,500
Q1	2015	\$1,000,000	0.0075	\$7,500
Q2	2015	\$1,000,000	0.0075	\$7,500

PROJECT ABSTRACT

**Figure 1**  
Trends in Self-Reported Happiness, 1971–1973



family context

work context

*Note.* Estimates are derived from sample surveys of noninstitutionalized population of the continental United States, aged 18 and over. Error bars demark  $\pm 1$  standard error around sample estimate. NORC = National Opinion Research Center; SRC = Survey Research Center. Questions were "Taken all together, how would you say things are these days—would you say that you are very happy, pretty happy, or not too happy?" (NORC); and "Taking all things together, how would you say things are these days—would you say you're very happy, pretty happy, or not too happy these days?" (SRC). From "Why Do Surveys Disagree? Some Preliminary Hypotheses and Some Disagreeable Examples" (p. 166) by C. F. Turner, 1984, in C. F. Turner and E. Martin, *Surveying Subjective Phenomena*, New York: Russell Sage Foundation. Copyright 1984 by the Russell Sage Foundation. Reprinted by permission.

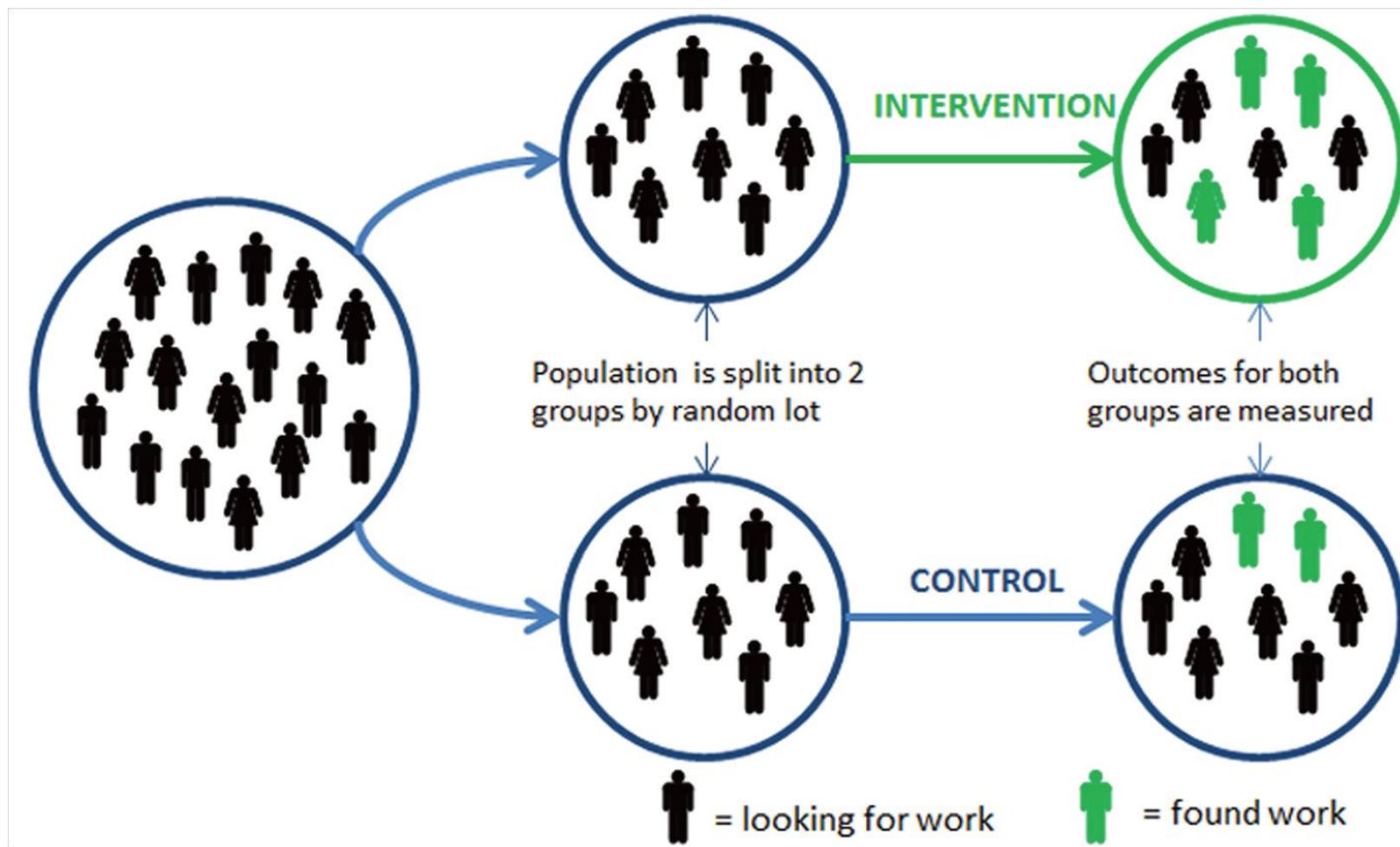
## causal methodologies

1. randomized controlled trials (RCTs)
2. “natural” experiments
3. pre-post comparisons
4. multiple regression / matching
5. instrumental variables
6. regression discontinuity

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2. “natural” experiments
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# RCTs





# body-worn camera study

[bwc.thelab.dc.gov](http://bwc.thelab.dc.gov)

## causal methodologies

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- 4. multiple regression / matching**
5. instrumental variables
6. regression discontinuity

The regression was significant, ( $R^2 = .05$ ),  $F(9, 1775) = 12.96$ ,  $p < .001$ . But the effect of language condition was marginal and had a negative rather than the expected positive slope,  $\beta = -.039$ , heteroskedasticity-consistent (HC) SE =  $.23$ ,  $p = .088$ .

Run MATRIX procedure:

HC Method  
0

Criterion Variable  
chose\_AB

Model Fit:

	R-sq	F	df1	df2	p
	.0545	12.9651	9.0000	1775.0000	.0000

Heteroskedasticity-Consistent Regression Results

	Coeff	SE(HC)	t	P> t
Constant	.4749	.0374	12.7038	.0000
language	-.0390	.0229	-1.7050	.0884
saw_scen	-.2057	.0476	-4.3266	.0000
saw_sc_1	-.0421	.0495	-.8503	.3953
saw_sc_2	.0309	.0507	.6093	.5424
saw_sc_3	.0243	.0503	.4829	.6292
saw_sc_4	-.0159	.0500	-.3190	.7498
saw_sc_5	-.1196	.0484	-2.4727	.0135
saw_sc_6	-.0122	.0498	-.2457	.8060
saw_sc_7	.2393	.0477	5.0122	.0000

Covariance Matrix of Parameter Estimates

	Constant	language	saw_scen	saw_sc_1	saw_sc_2	saw_sc_3	saw_sc_4	saw_sc_5
Constant	.0014	-.0003	-.0013	-.0013	-.0013	-.0013	-.0013	-.0013
language	-.0003	.0005	.0000	.0000	.0000	.0000	.0000	.0000
saw_scen	-.0013	.0000	.0023	.0013	.0013	.0013	.0013	.0013
saw_sc_1	-.0013	.0000	.0013	.0025	.0013	.0013	.0013	.0013
saw_sc_2	-.0013	.0000	.0013	.0013	.0026	.0013	.0013	.0013
saw_sc_3	-.0013	.0000	.0013	.0013	.0013	.0025	.0013	.0013
saw_sc_4	-.0013	.0000	.0013	.0013	.0013	.0013	.0025	.0013
saw_sc_5	-.0013	.0000	.0013	.0013	.0013	.0013	.0013	.0023
saw_sc_6	-.0012	.0000	.0013	.0013	.0013	.0013	.0013	.0013
saw_sc_7	-.0013	.0000	.0013	.0013	.0013	.0013	.0013	.0013

----- END MATRIX -----



# Regression

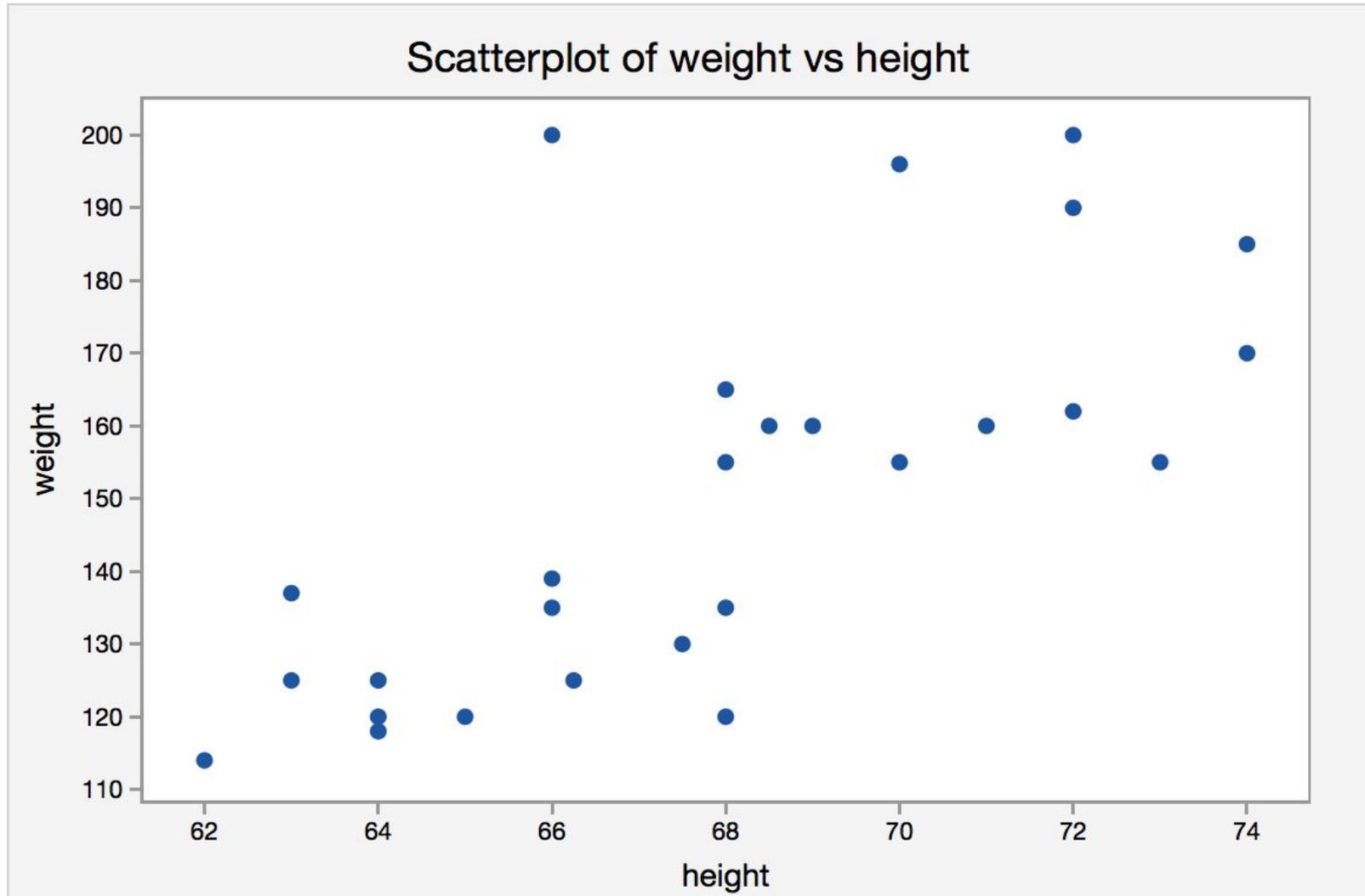
the basics

$$\hat{y}_i = a + b_1x_1$$

- $\text{Weight} = 115 + 8.6(\text{Height\_Inches})$ 
  - all else equal, each additional inch of height predicts an additional 8.6 pounds of weight

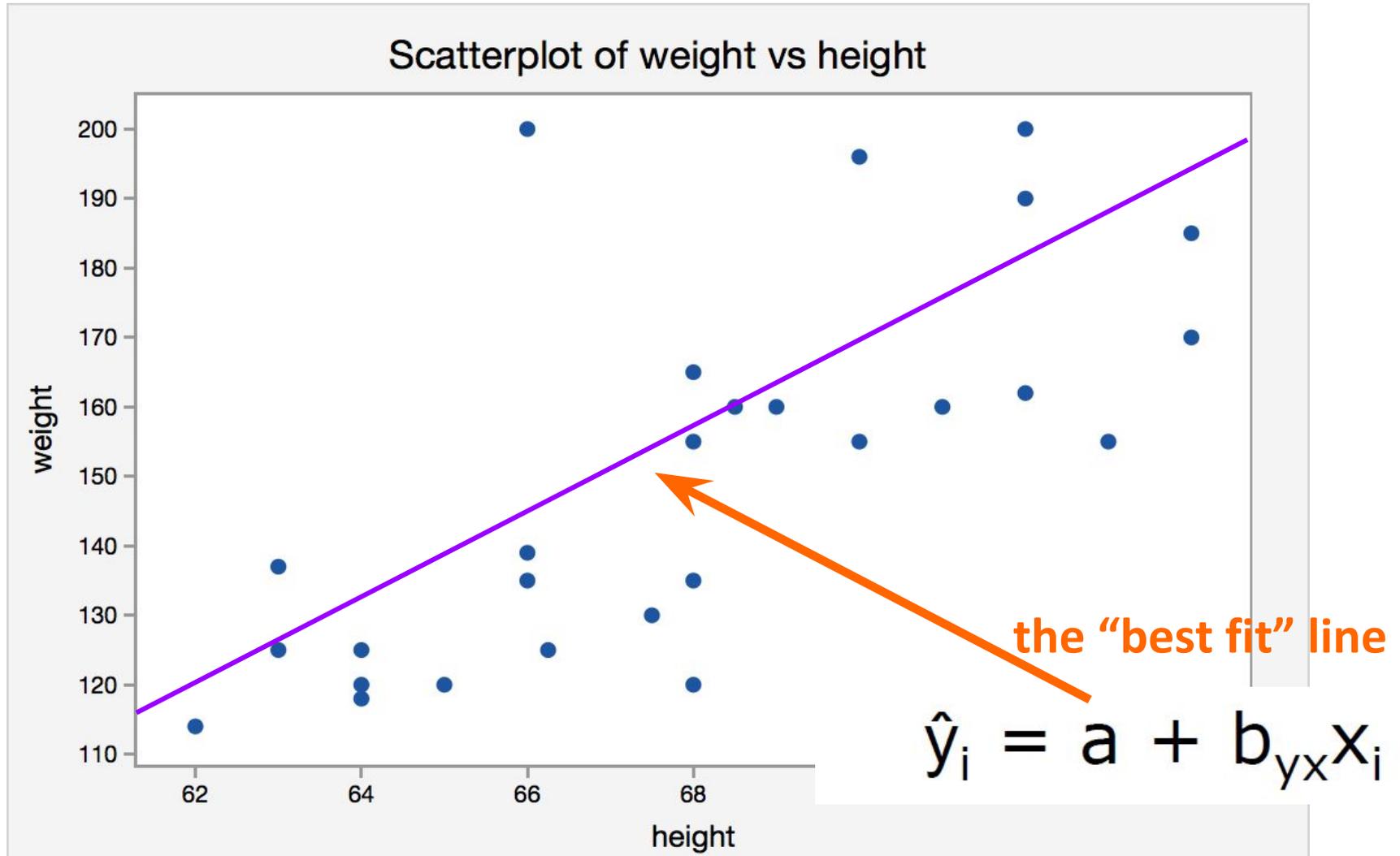
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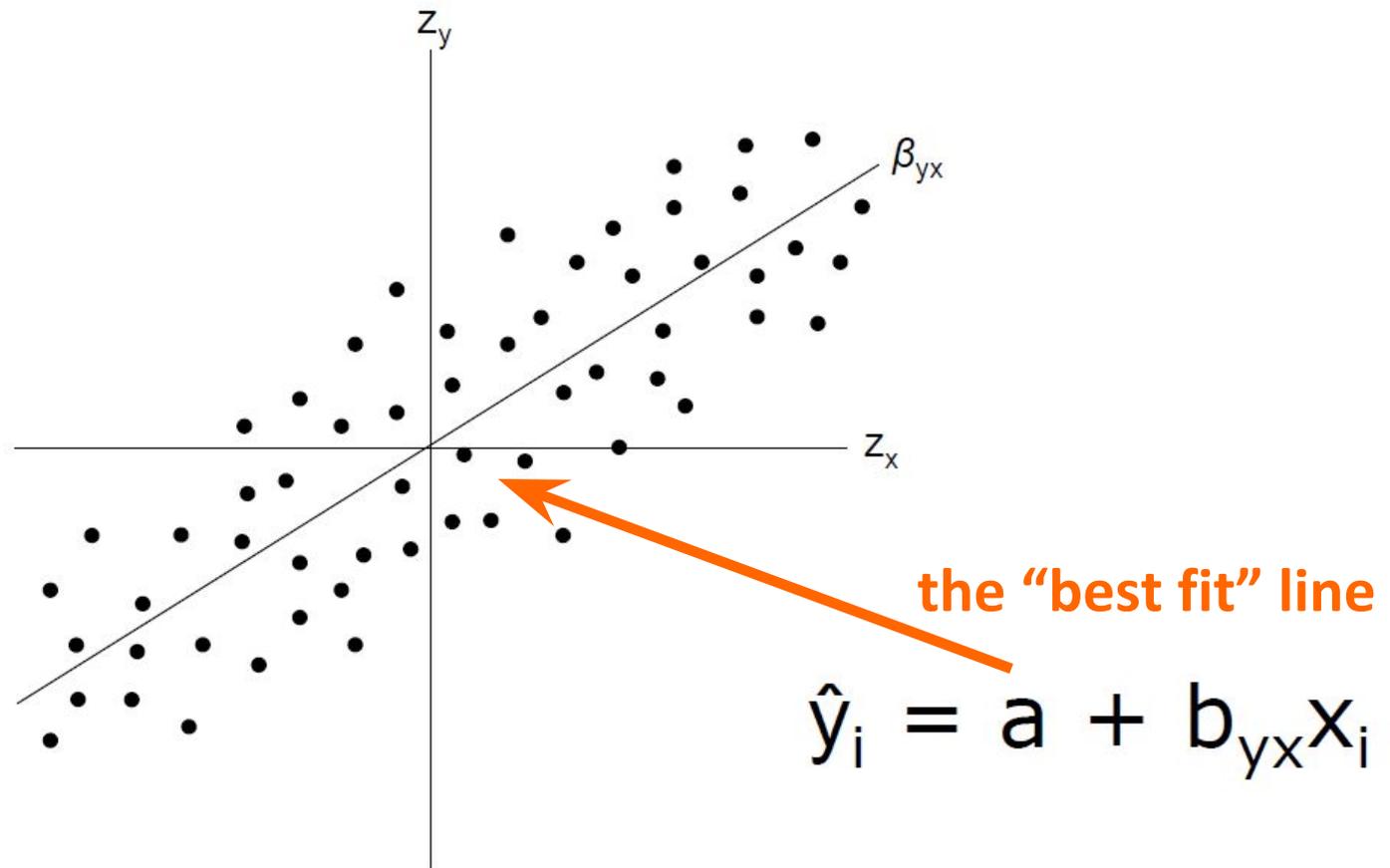
# Regression

the basics

- An interactive app, to get an intuitive feel for the “best fit” line:
  - <http://www.shodor.org/interactivate/activities/Regression/>

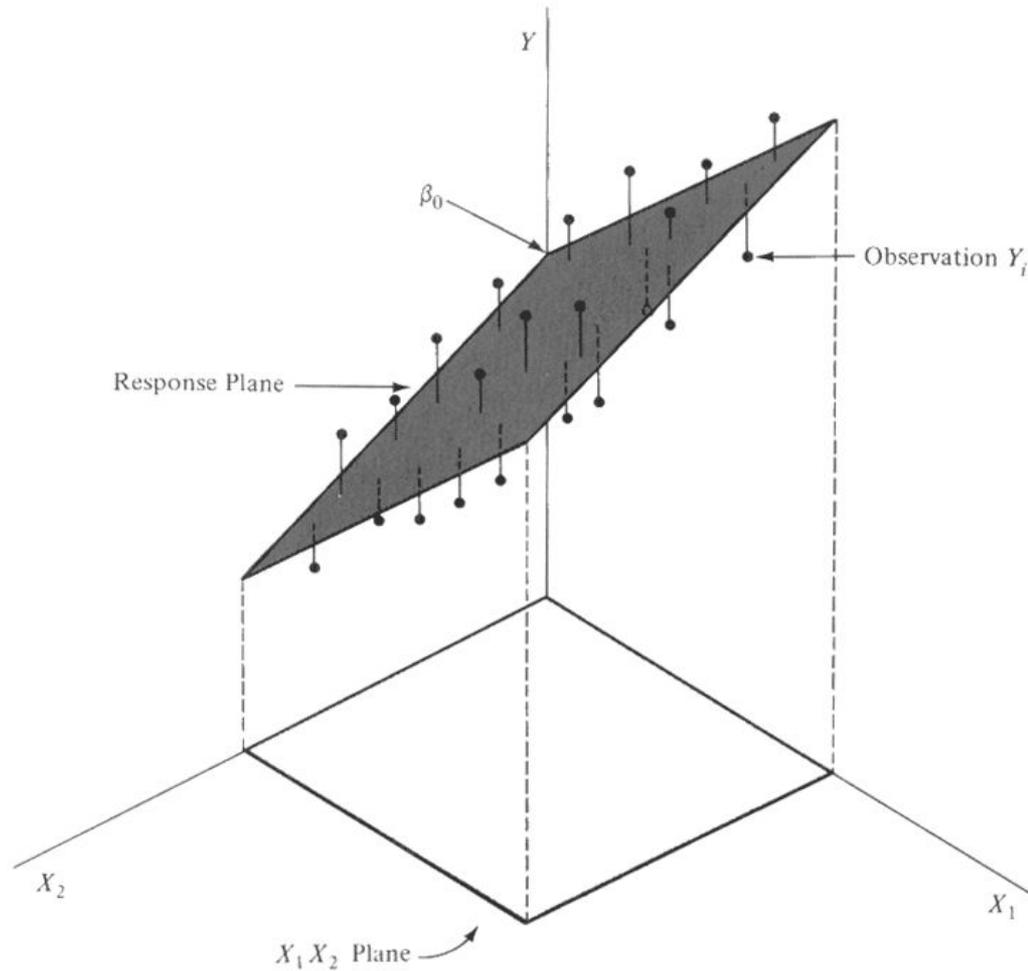
# Regression

visualizing one predictor



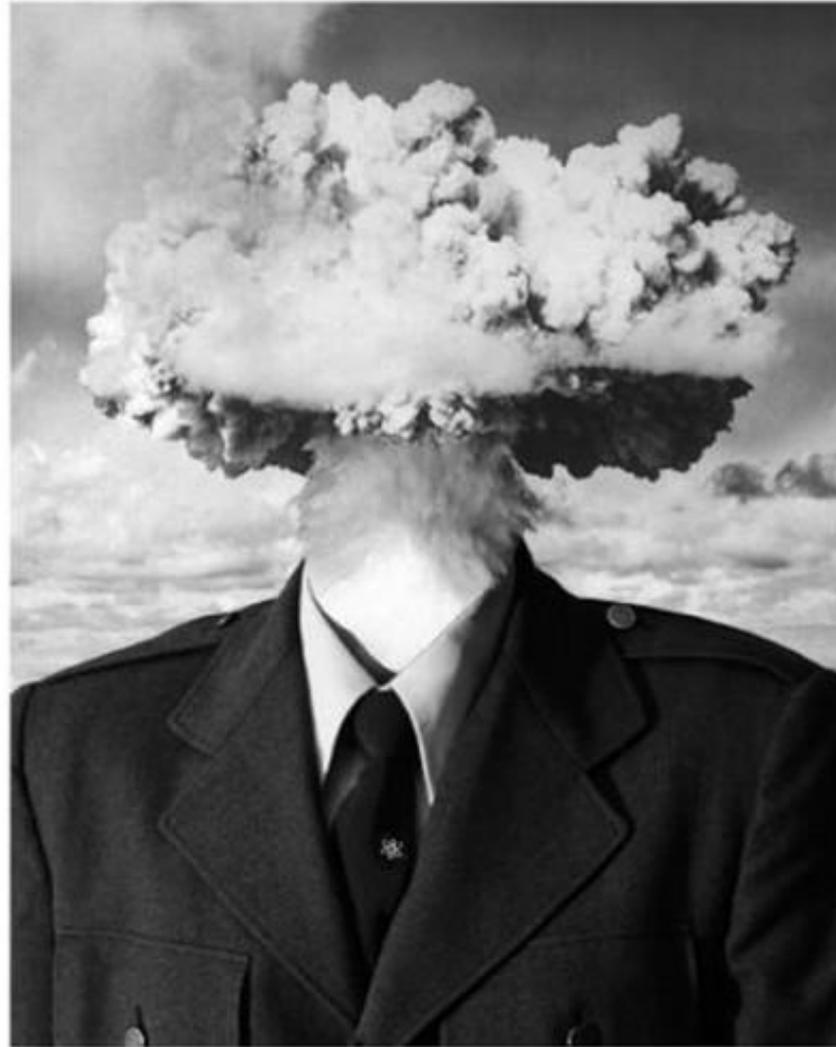
# Regression

visualizing two predictors



# Regression

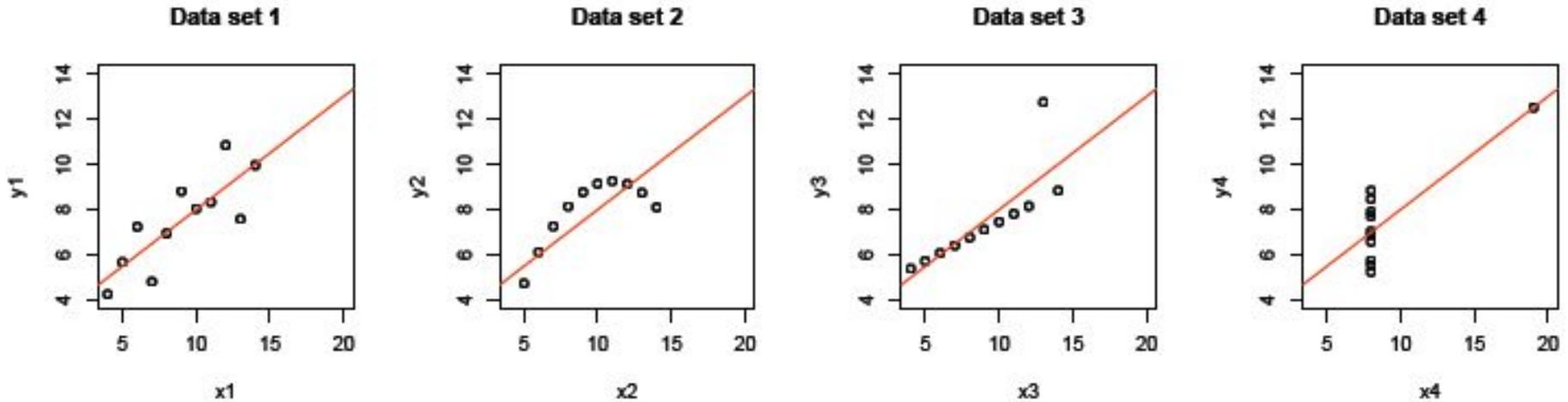
visualizing three plus predictors



# Regression

two tips

## 1. Look at the data.



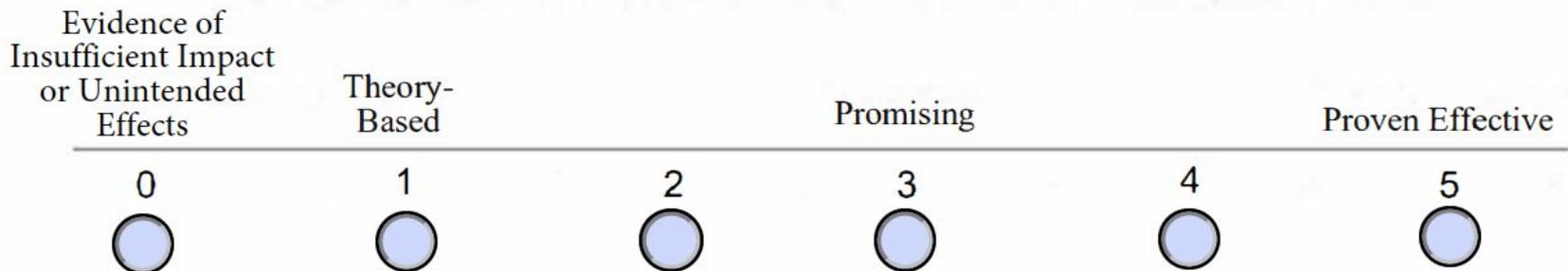
## 2. Ask yourself whether all the relevant predictors are included in the model.

the evidence scale

# an evidence continuum

## 6. Evidence Scale Ranking

*Please rank the proposed initiative's current level of evidentiary support on a scale from 0-5, based on the [RI Evidence Scale](#), with one being the least evidentiary support and five being the most evidentiary support. You can use tools like the [Pew Results First Clearinghouse](#) and the [Social Programs That Work](#) database to determine whether the initiative you are proposing has been rigorously evaluated in other jurisdictions. The Office of Management & Budget understands that the majority of agency requests will likely not be in the top evidence tiers at the point of submittal, and you should certainly feel free to submit requests that are "theory-based" rather than evidence based. Please note that "theory-based" submissions should include a robust and compelling measurement and evaluation plan in the Performance Measurement section.*



when and how should you  
generate more data?

# what don't you know, that matters?

<b>inputs</b>	<b>outputs</b>	<b>outcomes</b>	<b>impacts</b>
staff facilities materials etc.	# people served # job apps submitted	# employed average wages	increase employment or wages CAUSED BY the program

**implementation**

**outcome**

**impact**

## Forward Looking Opportunities to Develop an EvidenceBase

### 14. What methodologies **will you use** for program evaluation? Check all that apply.

- No evaluation is planned
- Measurement of the resources (e.g. staff, material expenses) required to deploy the service or initiative
- Measurement of how many people use the service or initiative
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- Other:

**what effect size(s)?**

# Opportunity: low-cost RCTs / evaluation

**11. Which of the following best describe the type(s) of data that you currently use for evaluation? Check all that apply.**

- No data is collected
- Qualitative data is collected (e.g. participant demographic information)
- Quantitative data is collected (e.g. number of participants)
- Aggregate-level data is collected (e.g. % of students who qualify for free/reduced lunch)
- Individual-level data is collected (e.g. household income amount for an individual student)

**12. Which of the following best describes the source(s) of data that you currently use for evaluation? Check all that apply.**

- No data is collected
- We use existing data that has been collected by another state agency, the federal government, a private entity, or another source
- We use existing data that has been collected by our agency for a different purpose
- We collect initiative-specific data

## basic tips to assess data quality

1. Look at your raw data
  - a. What's missing?
  - b. Do the entries make sense?
  - c. Do formats change over time?
2. Create a [data dictionary](#)
3. Do field work to understand how data were generated.
  - a. Pro Tip: Map it to your process map
4. Look at visualizations of your data
5. Compute basic descriptive statistics

## partnership opportunities

**18. Have you identified research partners to help evaluate the initiative? If so, please describe who. If not, what type of research partnership, if any, would be helpful?**

**19. Do you want to speak with a methods expert for a consultation on what evaluation methods might be best for learning about and optimizing the performance of your initiative?**

- No, we do not need a methods expert
- No, we have already engaged a methods expert
- Maybe, it might be useful, we're open to it
- Yes, that would be helpful

- Budget Instructions
  - Decision Package Template
  - Examples (good and bad)
  - Website Trainer
  - **Office Hours**
- 
- *Your feedback, much appreciated!*

**discussion**

## connect with The Policy Lab

- Explore at [thepolicylab.brown.edu](http://thepolicylab.brown.edu), and sign up for our listserv for updates, events, and ways to collaborate.
- Check out the podcast at [thirtythousandleagues.com](http://thirtythousandleagues.com).

