



# Why Randomize?

Dan Levy

Harvard Kennedy School





# Course Overview

1. What is Evaluation?
2. Outcomes, Impact, and Indicators
3. Why Randomize?
4. How to Randomize
5. Sampling and Sample Size
6. Threats and Analysis
7. Generalizability
8. Project from Start to Finish





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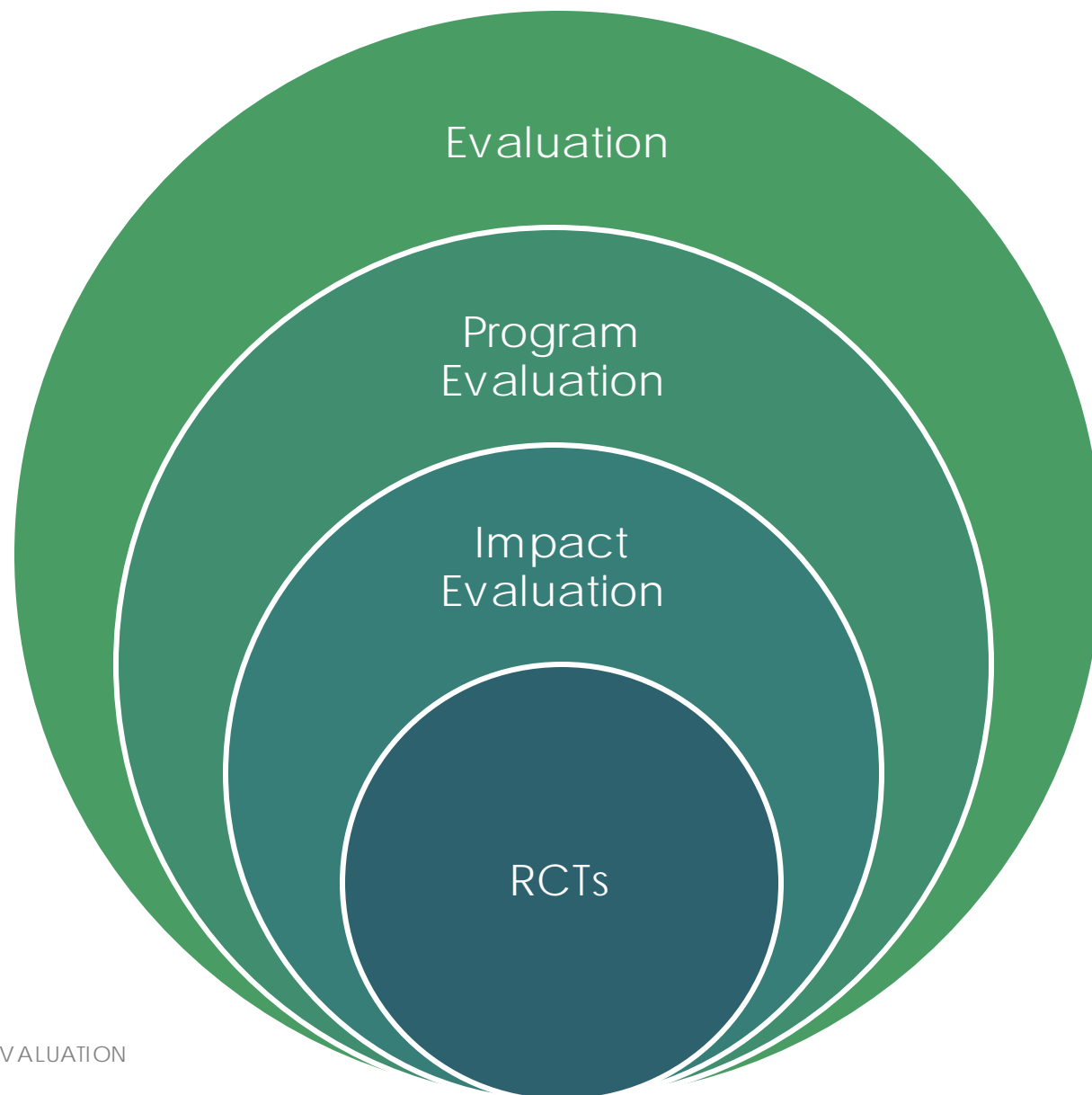


# Statistics Background





# What is Impact Evaluation ?



Methodologically, randomized controlled trials (RCTs) are the best approach to estimate the effect of a program

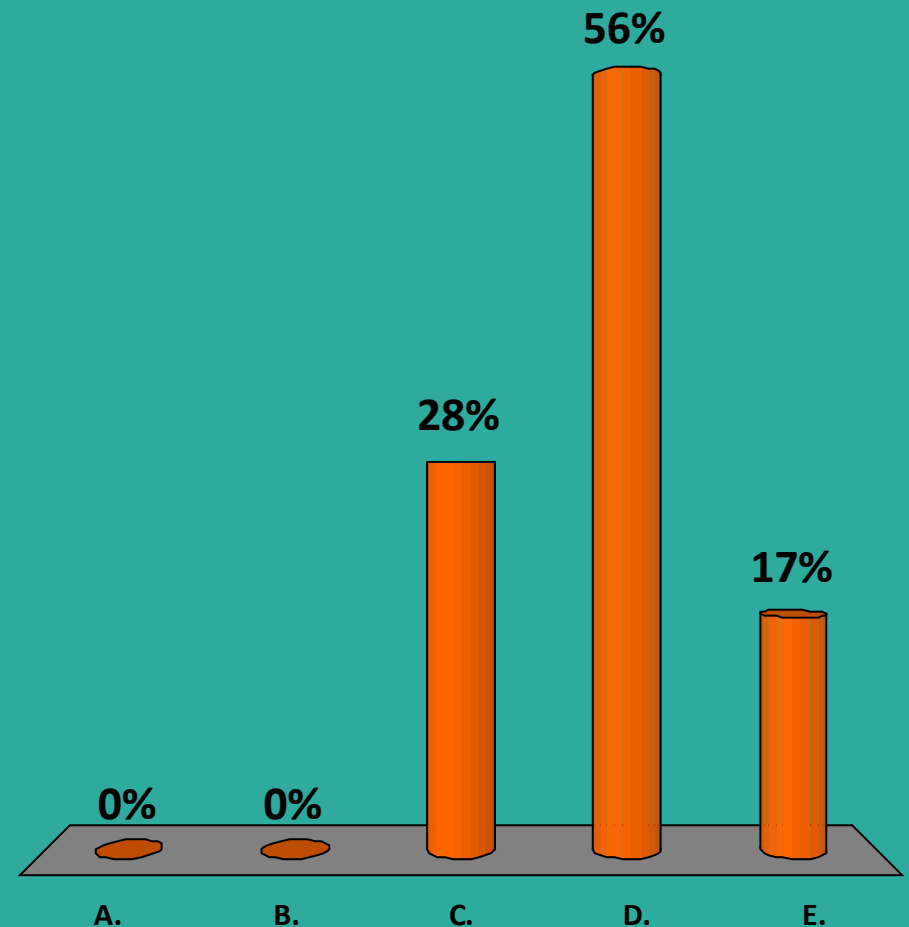
A. Strongly Disagree

B. Disagree

C. Neutral

D. Agree

E. Strongly Agree



# Session Overview

- I. Background
- II. What is a randomized experiment?
- III. Why randomize?
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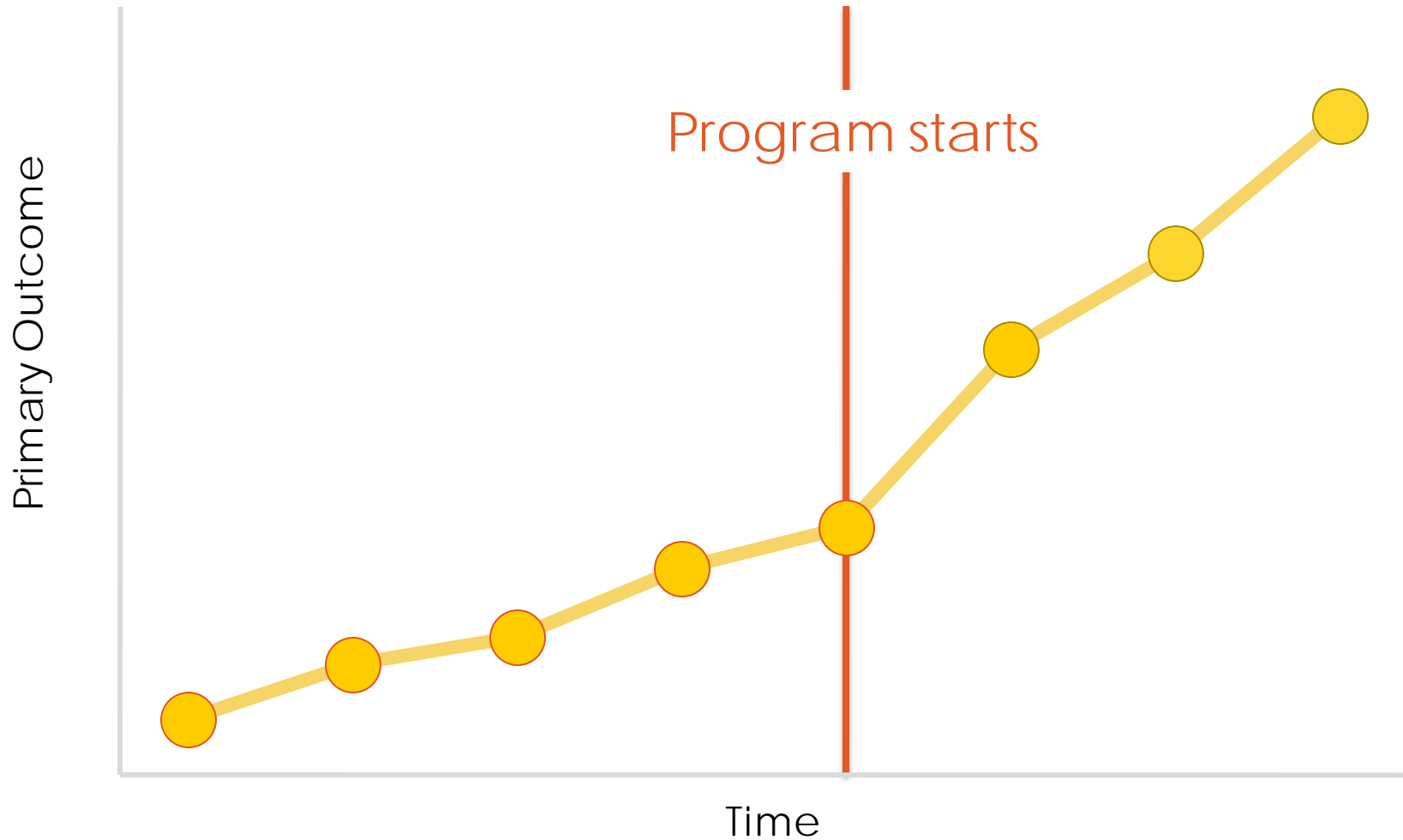


I - BACKGROUND



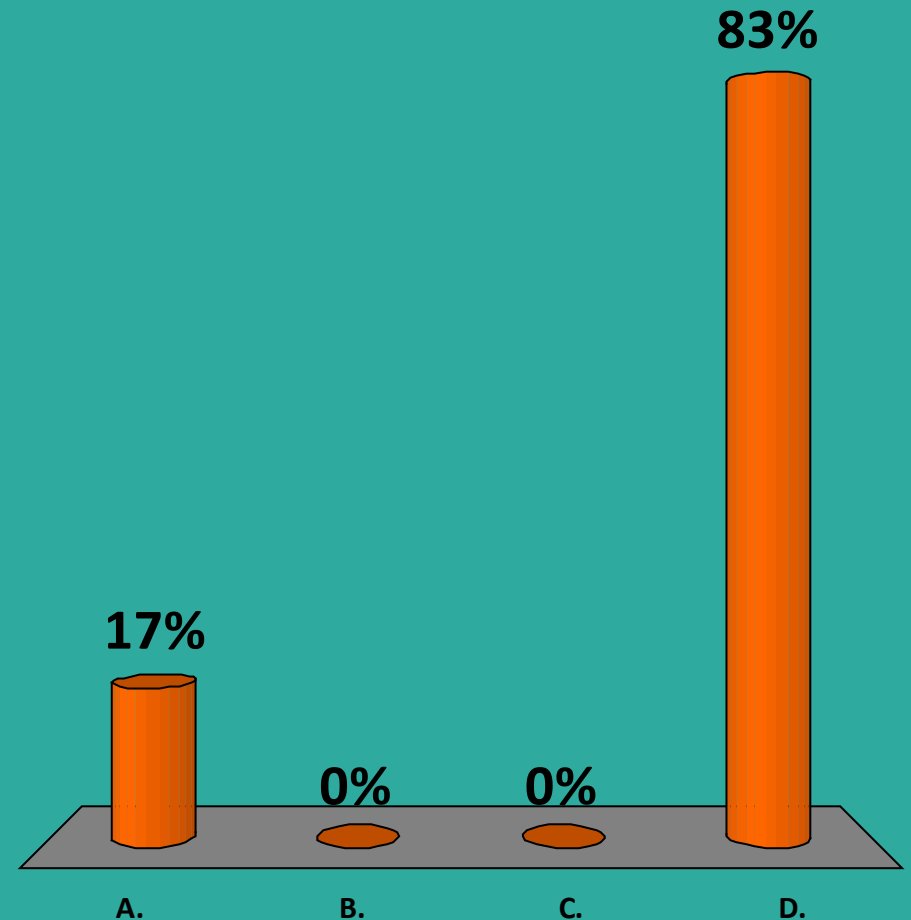


# What is the impact of this program?



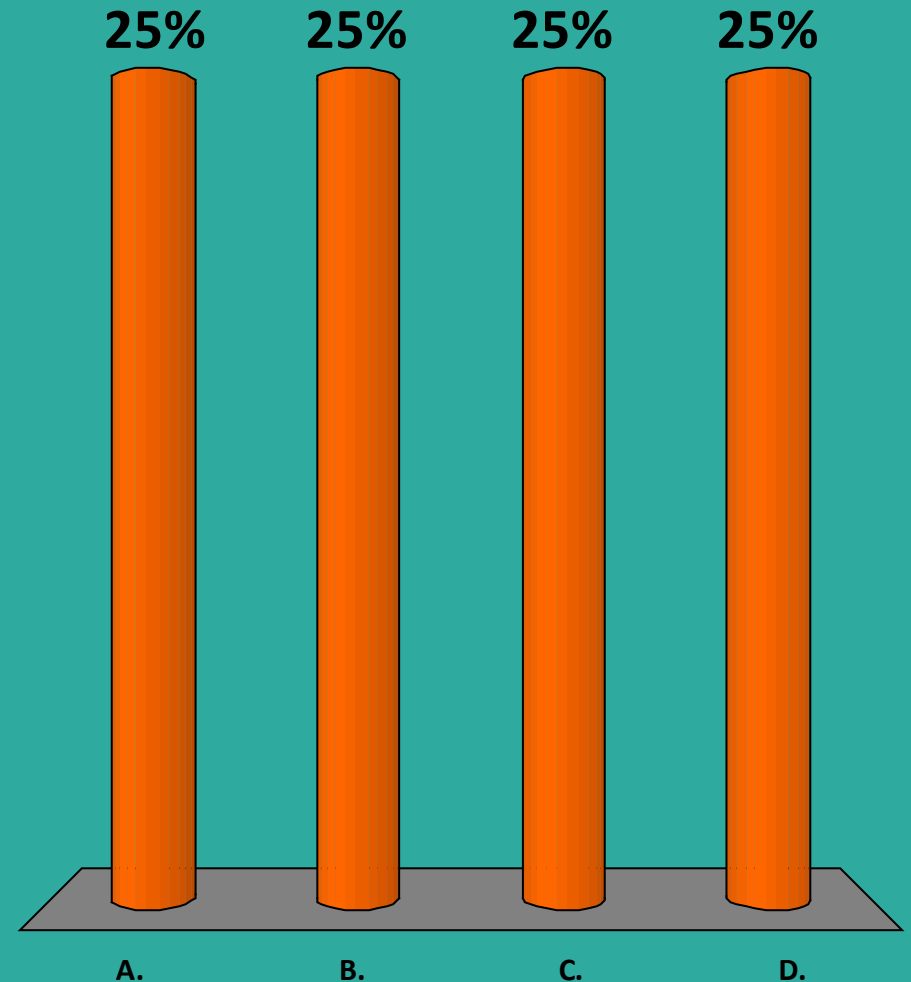
# What is the impact of this program?

- A. Positive
- B. Negative
- C. Zero
- D. Not enough info



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# Read India

FIGURE 1

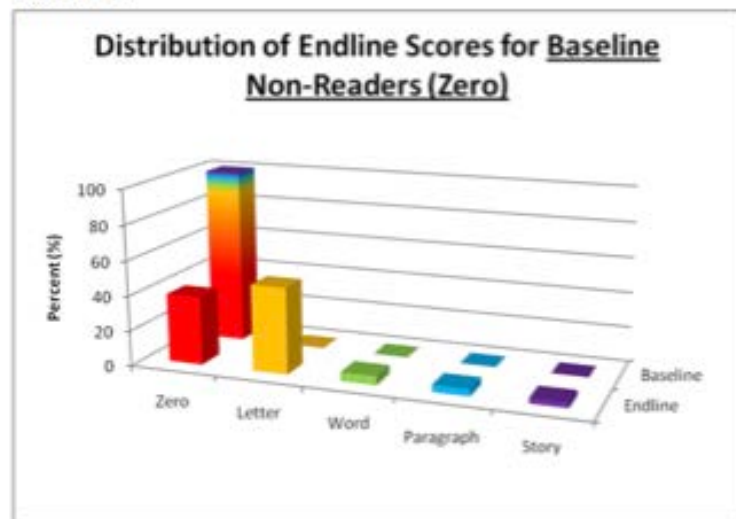
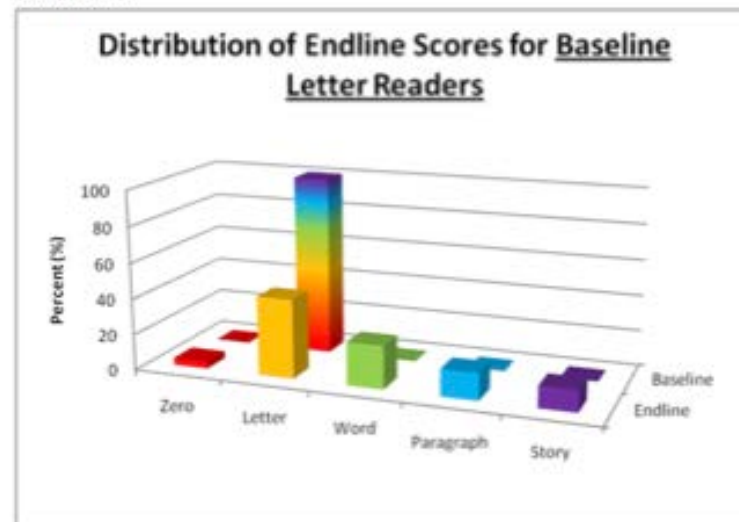
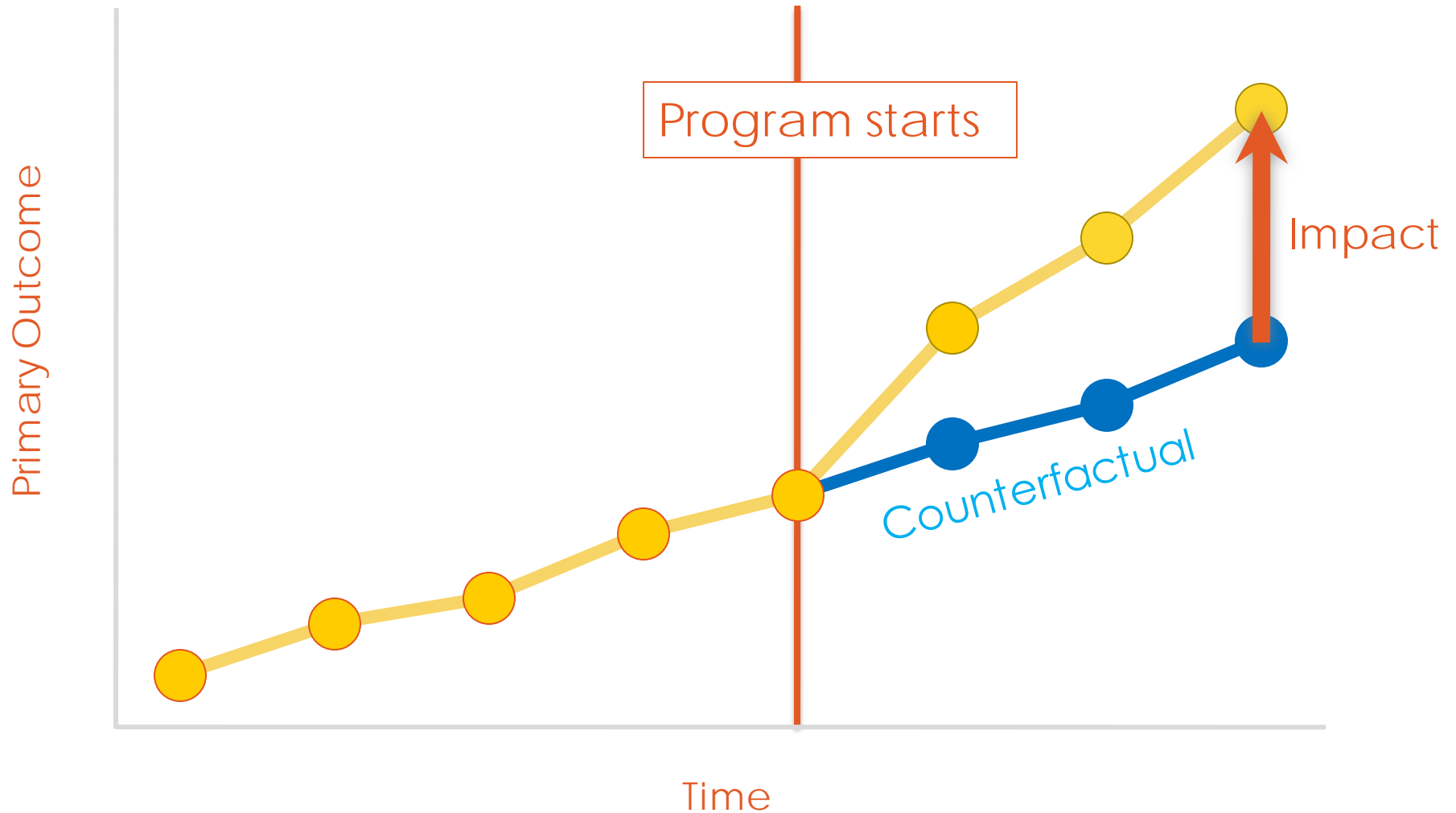


FIGURE 2



“Before vs. After” is rarely a good method for assessing impact.

# What is the impact of this program?





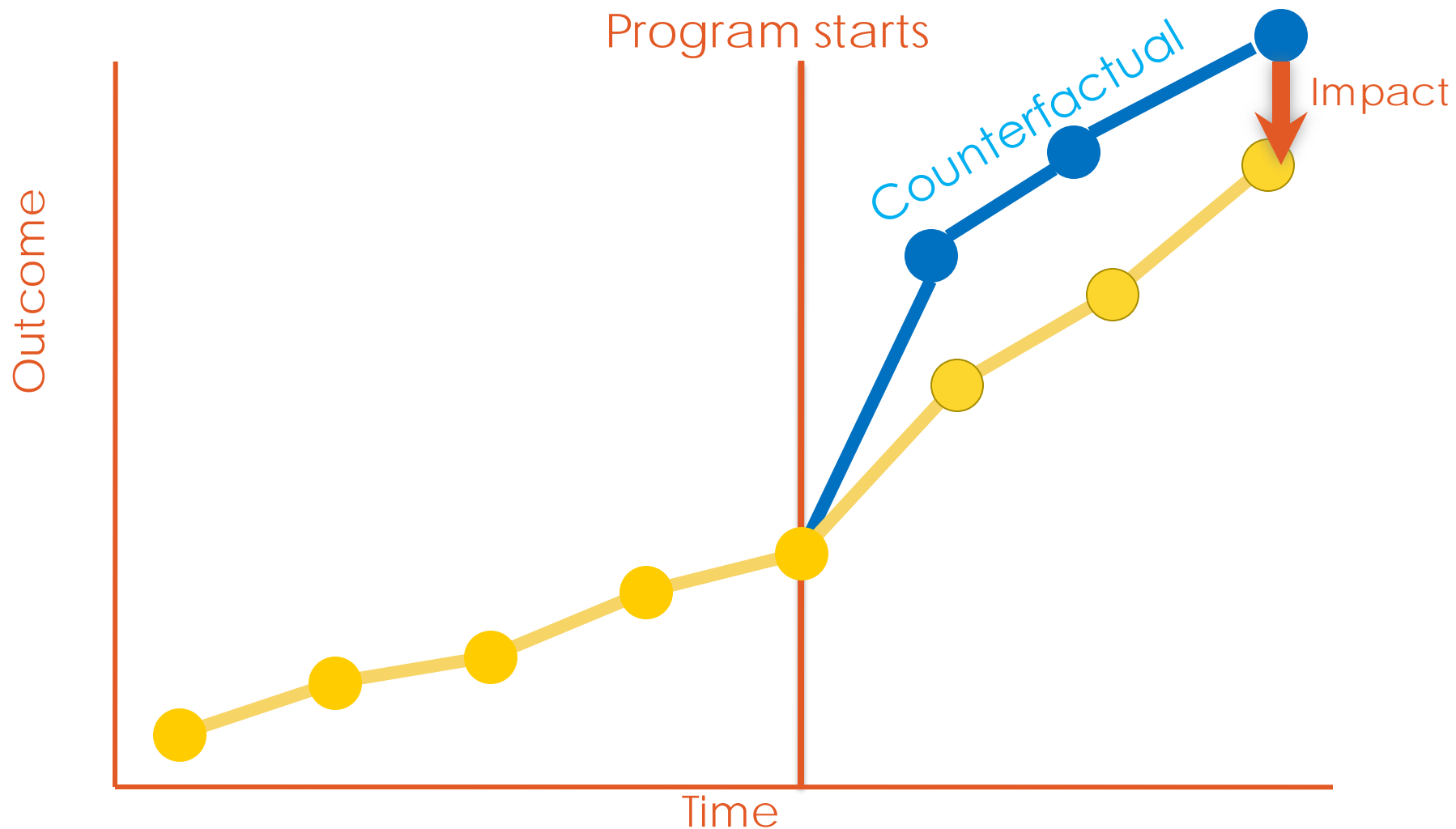
# How to measure impact?

- *Impact* is defined as a comparison between:
  1. the outcome some time after the program has been introduced (the "*factual*")
  2. the outcome at that same point in time had the program not been introduced (the "*counterfactual*")



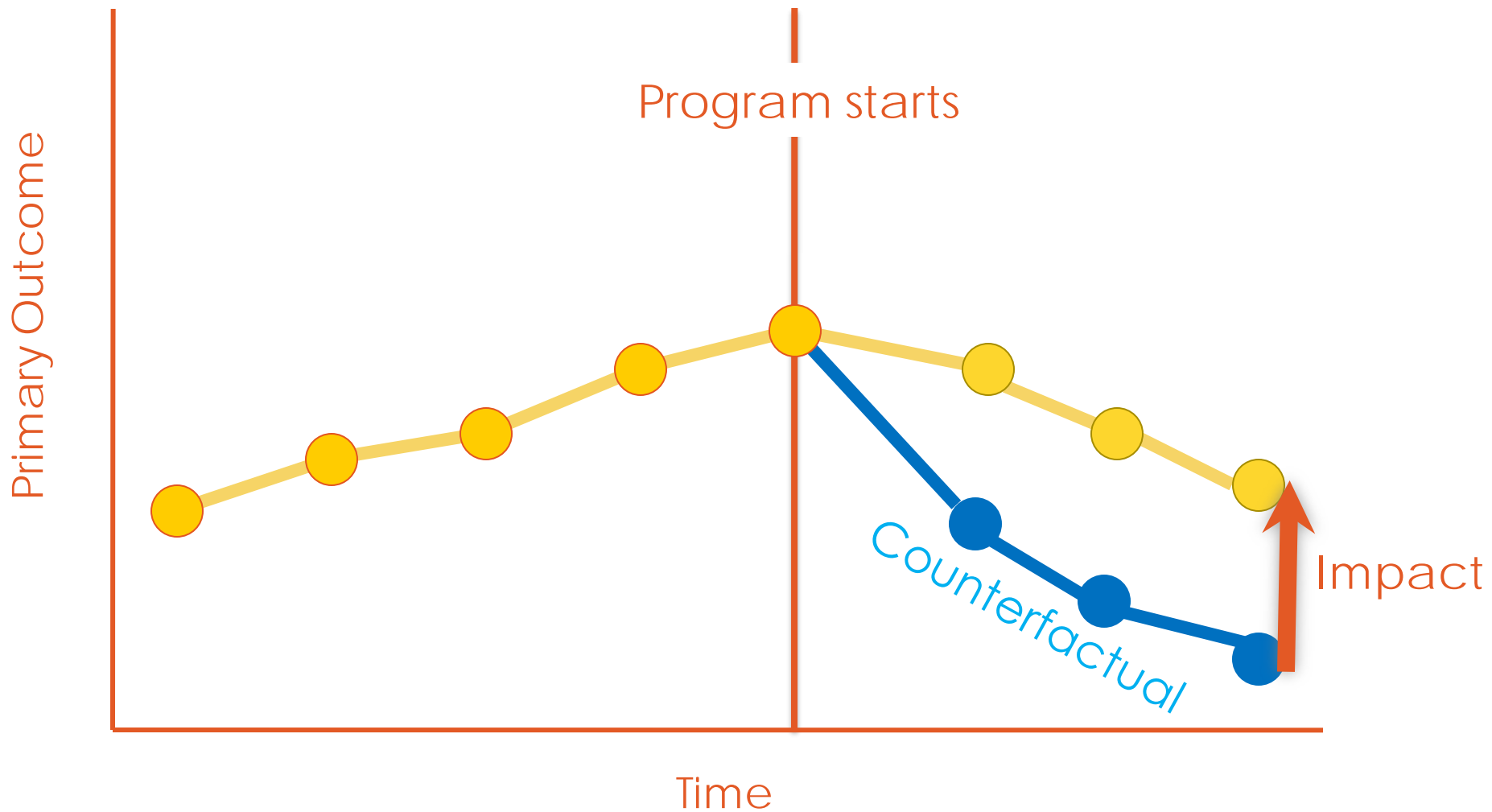


# Impact: What is it?





# Impact: What is it?





# Counterfactual

The *counterfactual* represents the state of the world that program participants would have experienced in the absence of the program

**Problem:** Counterfactual cannot be observed

**Solution:** We need to “mimic” or construct the counterfactual



# Constructing the counterfactual

- Usually done by selecting a group of individuals that **did not** participate in the program
- This group is usually referred to as the **control group** or **comparison group**
- How this group is selected is a **key decision** in the design of any impact evaluation

# Selecting the comparison group

- Idea: Comparability



- Goal: Attribution



# Impact evaluation methods

## 1. Randomized Controlled Trials (RCTs)

- Also known as:
  - Random Assignment Studies
  - Randomized Field Trials
  - Social Experiments
  - Randomized Trials
  - Randomized Experiments
  - Randomized Controlled Experiments

# Impact evaluation methods

## 2. Non- or Quasi-Experimental Methods

- a. Pre-Post
- b. Simple Difference
- c. Differences-in-Differences
- d. Multivariate Regression
- e. Statistical Matching
- f. Interrupted Time Series
- g. Instrumental Variables
- h. Regression Discontinuity

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## II – WHAT IS A RANDOMIZED EXPERIMENT?



# The basics

Start with simple case:

- Take a sample of program applicants
- *Randomly* assign them to either:
  - **Treatment Group** – is offered treatment
  - **Control Group** – not allowed to receive treatment (during the evaluation period)



# Key advantage of experiments

Because members of the groups (treatment and control) **do not differ systematically** at the outset of the experiment,

any difference that subsequently arises between them can be **attributed** to the program rather than to other factors.

# Evaluation of “Women as Policymakers”: Treatment vs. Control villages at baseline

Variables	Treatment Group	Control Group	Difference
Female Literacy Rate	0.35	0.34	0.01 (0.01)
Number of Public Health Facilities	0.06	0.08	-0.02 (0.02)
Tap Water	0.05	0.03	0.02 (0.02)
Number of Primary Schools	0.95	0.91	0.04 (0.08)
Number of High Schools	0.09	0.10	-0.01 (0.02)

Standard Errors in parentheses. Statistics displayed for West Bengal

\*/\*\*/\*\*\*: Statistically significant at the 10% / 5% / 1% level

Source: Chattopadhyay and Duflo (2004)



# Some variations on the basics

- Assigning to multiple treatment groups
- Assigning of units other than individuals or households
  - Health Centers
  - Schools
  - Local Governments
  - Villages



# Key Steps in conducting an experiment

1. **Design** the study carefully
2. **Randomly** assign people to treatment or control
3. Collect **baseline** data
4. **Verify** that assignment looks random
5. **Monitor** process so that integrity of experiments is not compromised



# Key Steps in conducting an experiment (contd.)

6. **Collect follow-up data** for both the treatment and control groups
7. Estimate program **impacts** by comparing mean outcomes of treatment group vs mean outcomes of the control group
8. Assess whether program impacts are **statistically significant** and **practically significant**

EXAMPLE







# Program “Get Out the Vote”

- Low voter turnout is seen as a problem in many countries in the world
- Some countries have looked for ways to increase voter turnout
- “Get Out the Vote” Program
  - Compiled a list of all the 100,000 individuals who could vote in an election
  - Call a sample individuals in this list
  - In this phone call, responder is encouraged to vote



# Program "Get Out the Vote"

)

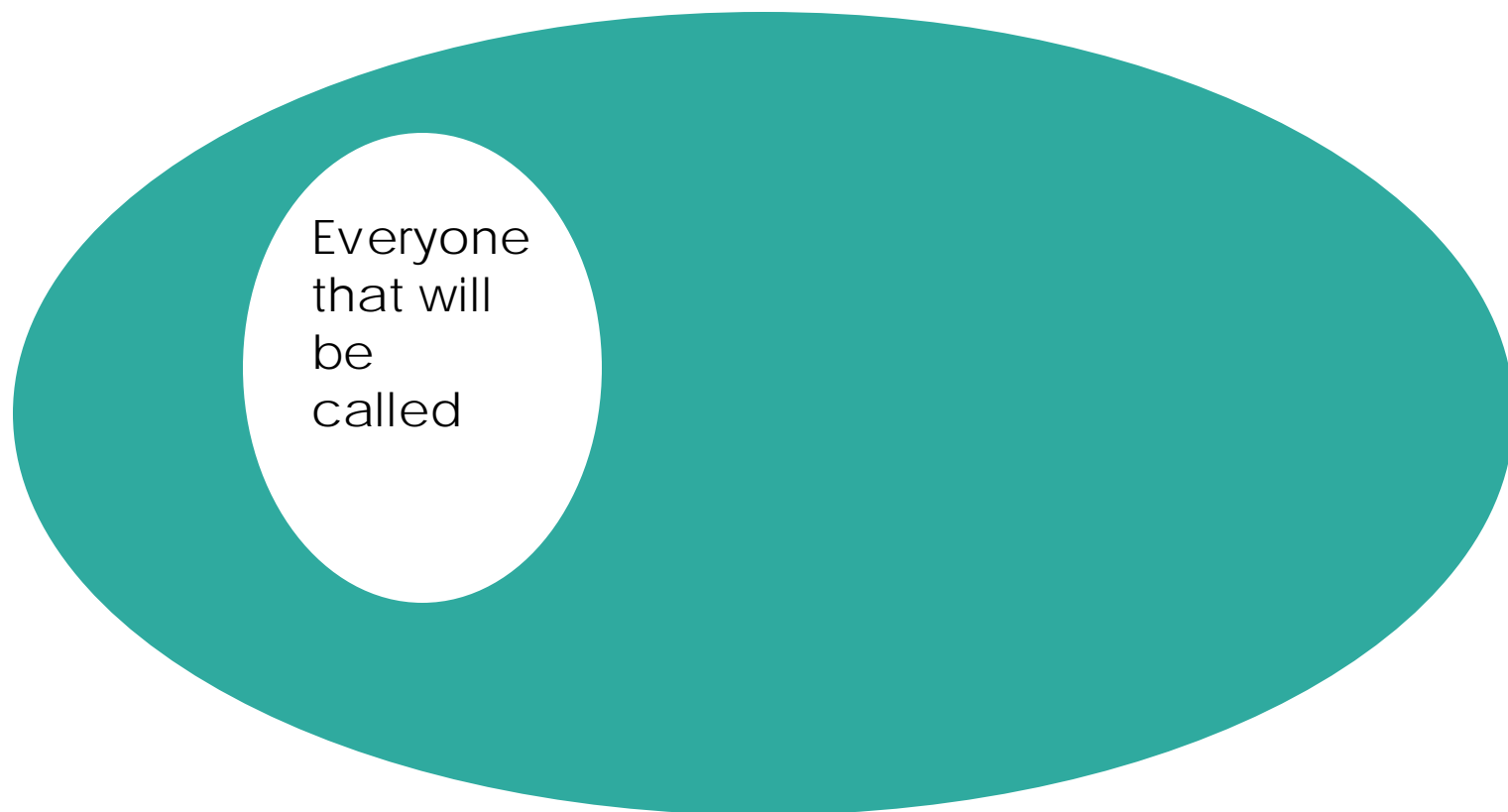
Everyone eligible to vote (100,000)





# Program "Get Out the Vote"

Everyone eligible to vote (100,000)





## Program “Get Out the Vote” (Contd.)

**Key question:** What is the **impact** of the “Get Out the Vote” program on the voter turnout rate?

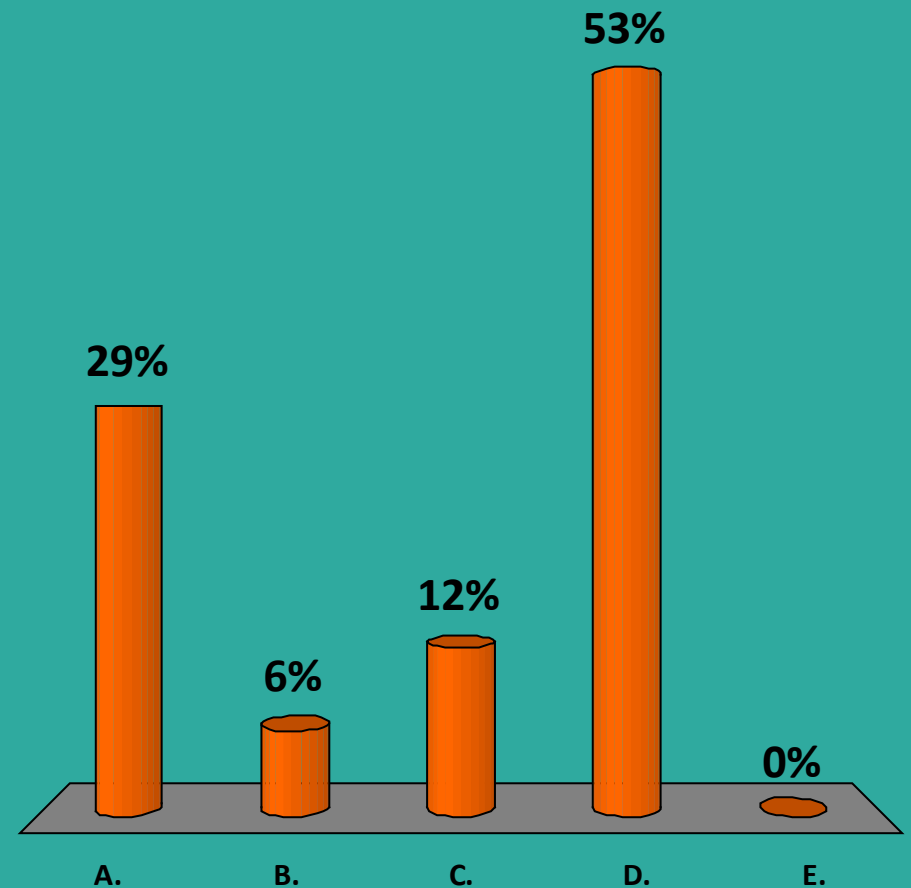
**Methodological Question:** How should we estimate the impact of the program?

# Resources available for the evaluation

- List of all the persons eligible to vote with information on:
  - Income
  - Education
  - Sex
  - Age
  - Whether person voted in the last election
- Money to make up to 8,000 calls that could be used to:
  - Implement the program (i.e. call before the election encouraging person to vote)
  - Collect data (i.e. call people after the election to ask whether they voted or not)
- List of 2,000 people who came to a political rally one month before the election
  - You already called them and encouraged them to vote
  - These calls count as part of your 8,000 calls

# Which design would you choose?

- A. Design 1
- B. Design 2
- C. Design 3
- D. Design 4
- E. Design 5



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### III – WHY RANDOMIZE?







# Why Randomize?- Conceptual Argument

If properly designed and conducted, randomized experiments provide the **most credible** method to estimate the impact of a program



# Why “most credible”?

Because members of the groups (treatment and control) **do not differ systematically** at the outset of the experiment,

any difference that subsequently arises between them can be **attributed** to the program rather than to other factors.

# Example #1 - Pratham's Read India program



# Example #1 - Pratham's Read India program

Method	Impact
(1) Pre-Post	0.60*
(2) Simple Difference	-0.90*
(3) Difference-in-Differences	0.31*
(4) Regression	0.06
(5) Randomized Experiment	

\*: Statistically significant at the 5% level

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<b>(5) Randomized Experiment</b>	<b>0.88*</b>

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# Example #2: A voting campaign in the USA



Courtesy of Flickr user theocean

# A voting campaign in the USA

Method	Impact (Vote %)
(1) Pre-Post	-7.2 pp
(2) Simple Difference	10.8 pp*
(3) Difference-in-Differences	3.8 pp*
(4) Multiple Regression	6.1 pp*
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Bottom Line: Which method we use matters

## IV – CONCLUSIONS

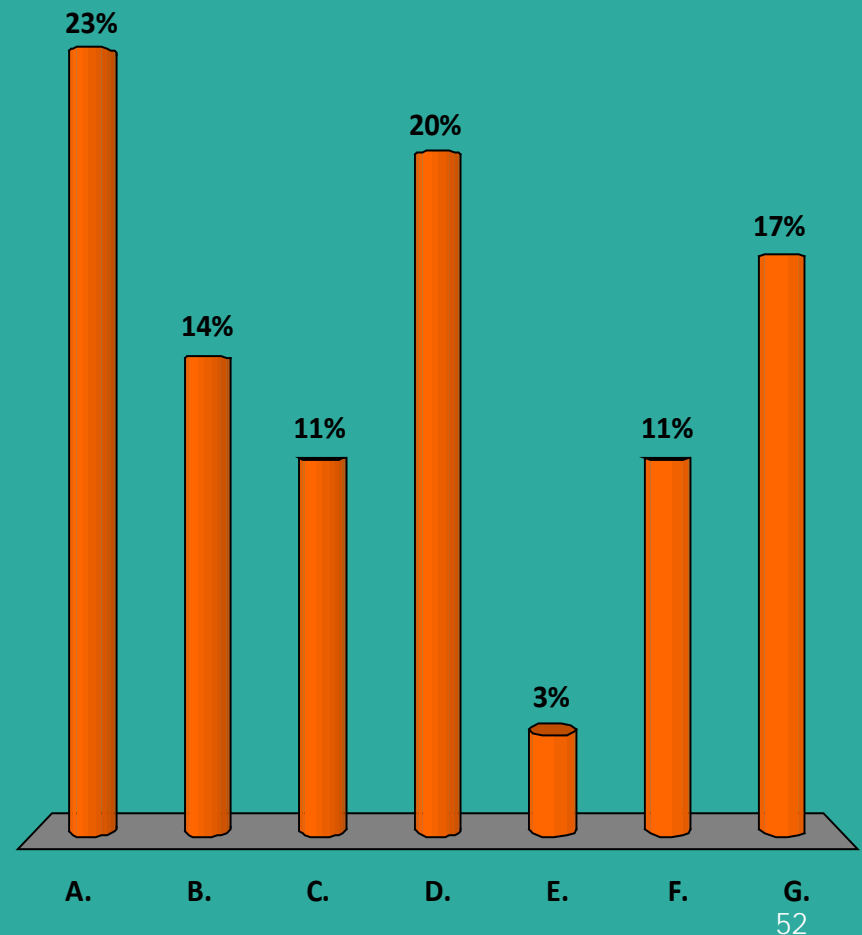


# Conclusions - Why Randomize?

- There are **many ways** to estimate a program's impact
- This course argues in favor of one: **randomized experiments**
  - **Conceptual argument:** If properly designed and conducted, randomized experiments provide the most credible method to estimate the impact of a program
  - **Empirical argument:** Different methods can generate different impact estimates

# What is the most convincing argument you have heard against RCTs? Enter your **top 3 choices**.

- A. Too expensive
- B. Takes too long
- C. Not ethical
- D. Too difficult to design/implement
- E. Not externally valid (Not generalizable)
- F. Less practical to implement than other methods and not much better
- G. Can tell us *what the impact is* impact, but not *why* or *how* it occurred (i.e. it is a black box)

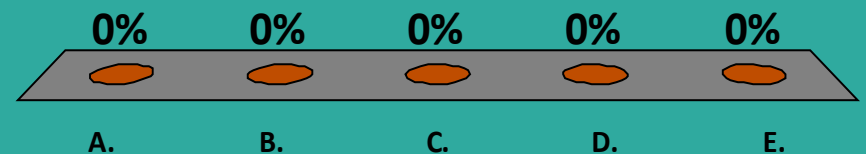


THANK YOU!



Methodologically, randomized trials are the best approach to estimate the effect of a program

- A. Strongly Disagree
- B. Disagree
- C. Neutral
- D. Agree
- E. Strongly Agree





J-PAL

ABDUL LATIF JAMEEL POVERTY ACTION LAB

# Why Randomize? Backup Slides

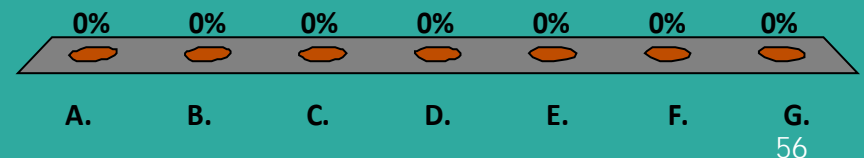
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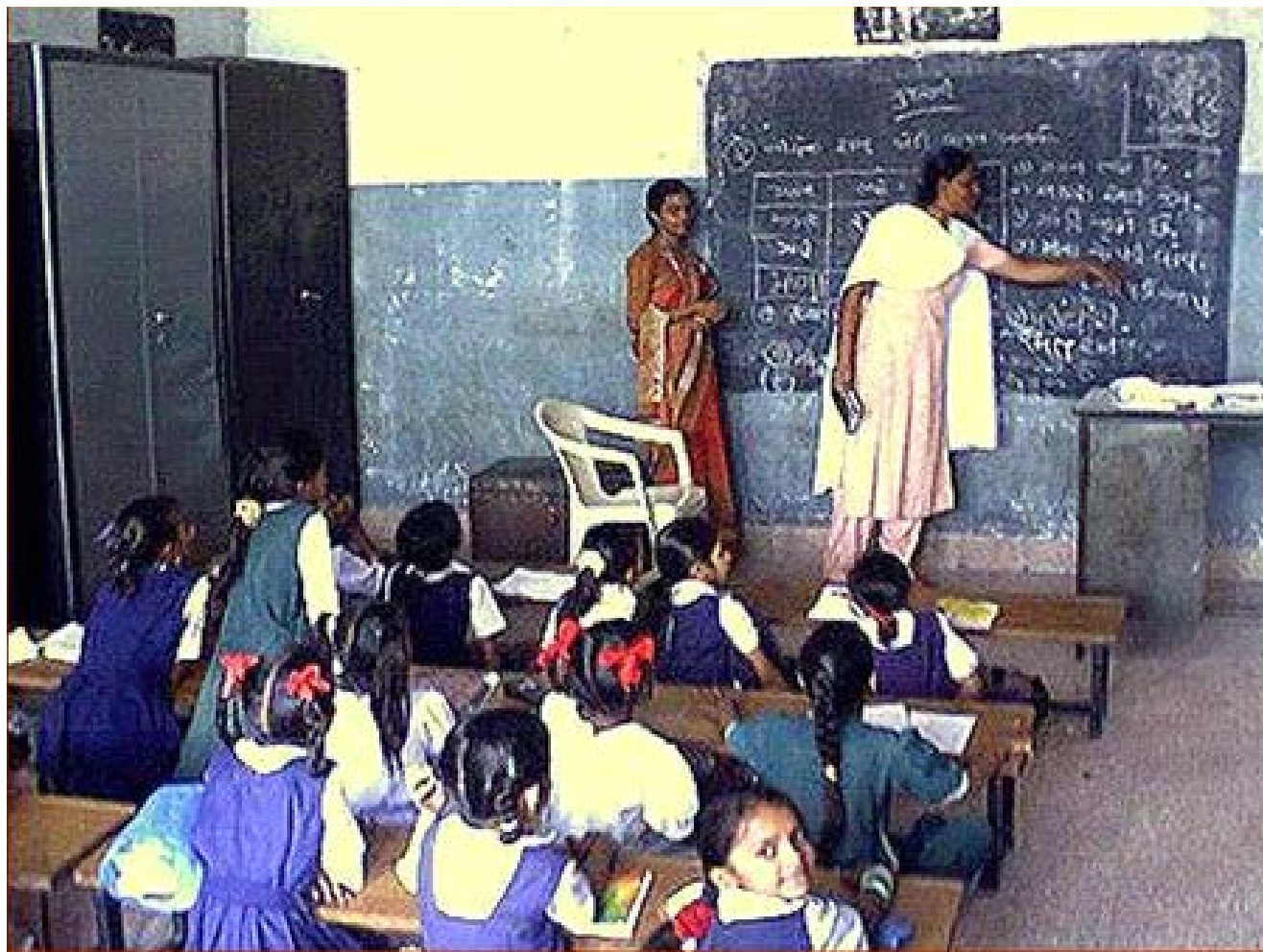


# What do you want to do?

- A. Example
- B. Objections to RCTs



# Example #3 Balsakhi Program





# Balsakhi Program: Background

- Implemented by **Pratham**, an NGO from India
- Program provided **tutors** ( Balsakhi) to help at-risk children with school work
- In Vadodara, the balsakhi program was run in government primary schools in **2002-2003**
- **Teachers decided** which children would get the balsakhi

# Balsakhi: Outcomes

- Children were tested at the beginning of the school year (Pretest) and at the end of the year (Post-test)
- **QUESTION:** How can we estimate the impact of the balsakhi program on test scores?

# Methods to estimate impacts

- Let's look at different ways of estimating the impacts using the data from the schools that got a balsakhi
  1. Pre – Post (Before vs. After)
  2. Simple difference
  3. Difference-in-difference
  4. Other non-experimental methods
  5. Randomized Experiment

# 1 - Pre-post (Before vs. After)

- Look at average change in test scores over the school year for the balsakhi children



# 1 - Pre-post (Before vs. After)

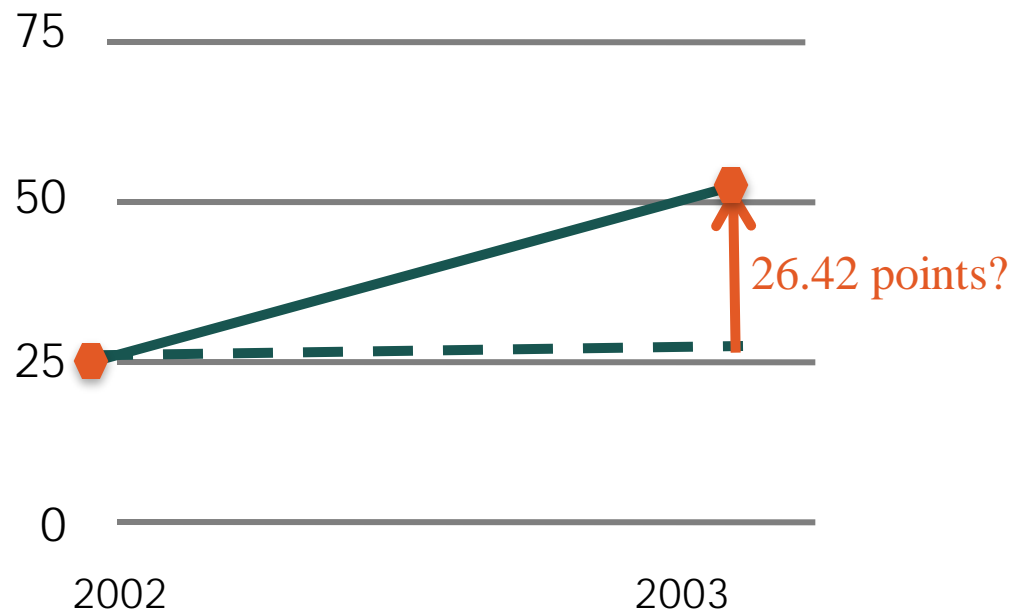
Average <u>post-test</u> score for children with a balsakhi	51.22
Average <u>pretest</u> score for children with a balsakhi	24.80
Difference	26.42

**QUESTION:** Under what conditions can this difference (26.42) be interpreted as the impact of the balsakhi program?

# What would have happened without balsakhi?

Method 1: Before vs. After

Impact = 26.42 points?





## 2 - Simple difference

Compare test scores of...



With  
test  
scores  
of...



Children who **got**  
balsakhi

Children who **did not**  
get balsakhi

## 2 - Simple difference

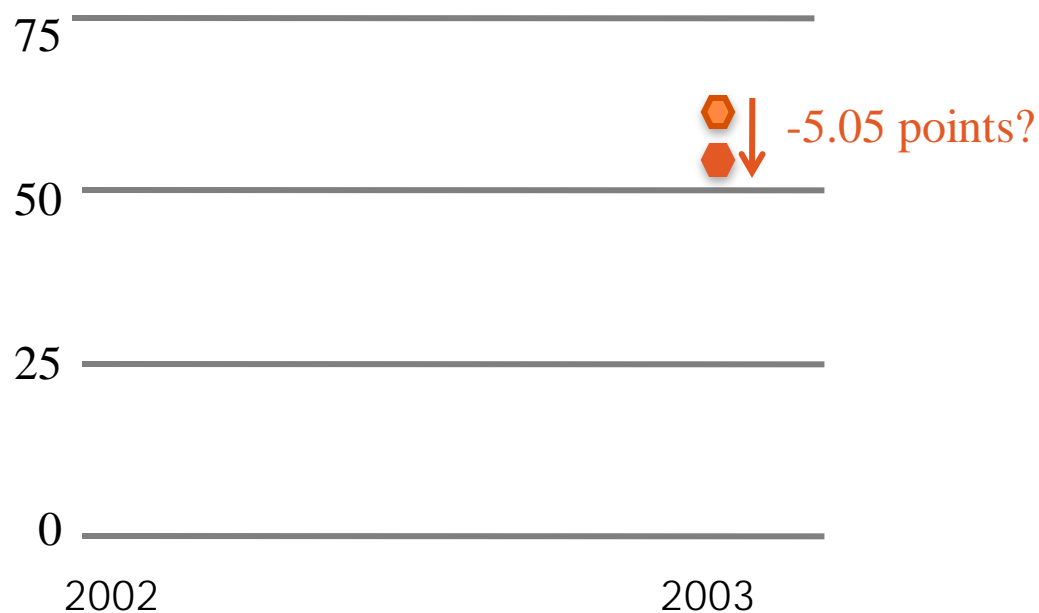
Average score for children with a balsakhi	51.22
Average score for children without a balsakhi	56.27
Difference	-5.05

**QUESTION:** Under what conditions can this difference (-5.05) be interpreted as the impact of the balsakhi program?

# What would have happened without balsakhi?

Method 2: Simple Comparison

Impact = -5.05 points?



# 3 – Difference-in-Differences

Compare gains in test scores of...



With gains in test scores of...



Children who **got** balsakhi

Children who **did not** get balsakhi

### 3 – Difference-in- difference

	Pretest	Post-test	Difference
Average score for children <b>with</b> a balsakhi	24.80	51.22	26.42

- **QUESTION:** Under what conditions can this difference (-5.05) be interpreted as the impact of the balsakhi program?

### 3 – Difference-in-difference

	Pretest	Post-test	Difference
Average score for children <b>with</b> a balsakhi	24.80	51.22	26.42
Average score for children <b>without</b> a balsakhi	36.67	56.27	19.60

### 3 – Difference-in-difference

	Pretest	Post-test	Difference
Average score for children <b>with</b> a balsakhi	24.80	51.22	26.42
Average score for children <b>without</b> a balsakhi	36.67	56.27	19.60
<b>Difference</b>			<b>6.82</b>

## 4 – Other Methods

- There are more sophisticated non-experimental methods to estimate program impacts:
  - Regression
  - Matching
  - Instrumental Variables
  - Regression Discontinuity
- These methods rely on being able to “mimic” the counterfactual **under certain assumptions**
- **Problem:** Assumptions are not testable



# 5 – Randomized Experiment

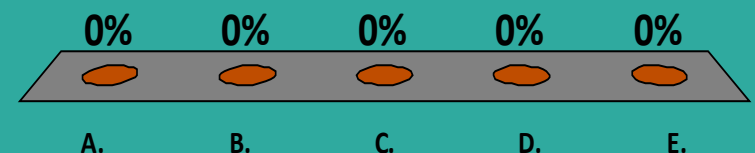
- Suppose we evaluated the balsakhi program using a randomized experiment
- **QUESTION #1:** What would this entail? How would we do it?
- **QUESTION #2:** What would be the advantage of using this method to evaluate the impact of the balsakhi program?

# Which of these methods do you think is closest to the truth?

Method	Impact Estimate
(1) Pre-post	26.42*
(2) Simple Difference	-5.05*
(3) Difference-in-Difference	6.82*
(4) Regression	1.92

\*: Statistically significant at the 5% level

- A. Pre-Post
- B. Simple Difference
- C. Difference-in-Differences
- D. Regression
- E. Don't know



# Impact of Balsakhi - Summary

Method	Impact Estimate
(1) Pre-Post	26.42*
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Bottom Line: Which method we use matters!

# Example #2 - Pratham's Read India program



## Example #2 – Pratham’s Read India program

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Courtesy of Flickr user theocan



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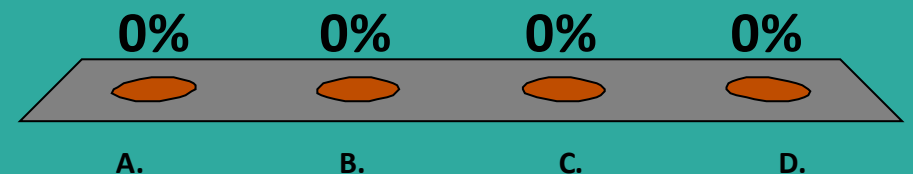
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THANK YOU!



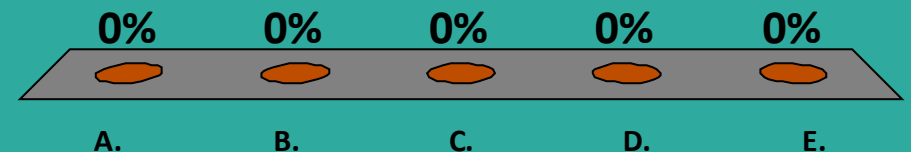
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- A. Positive
- B. Negative
- C. Zero
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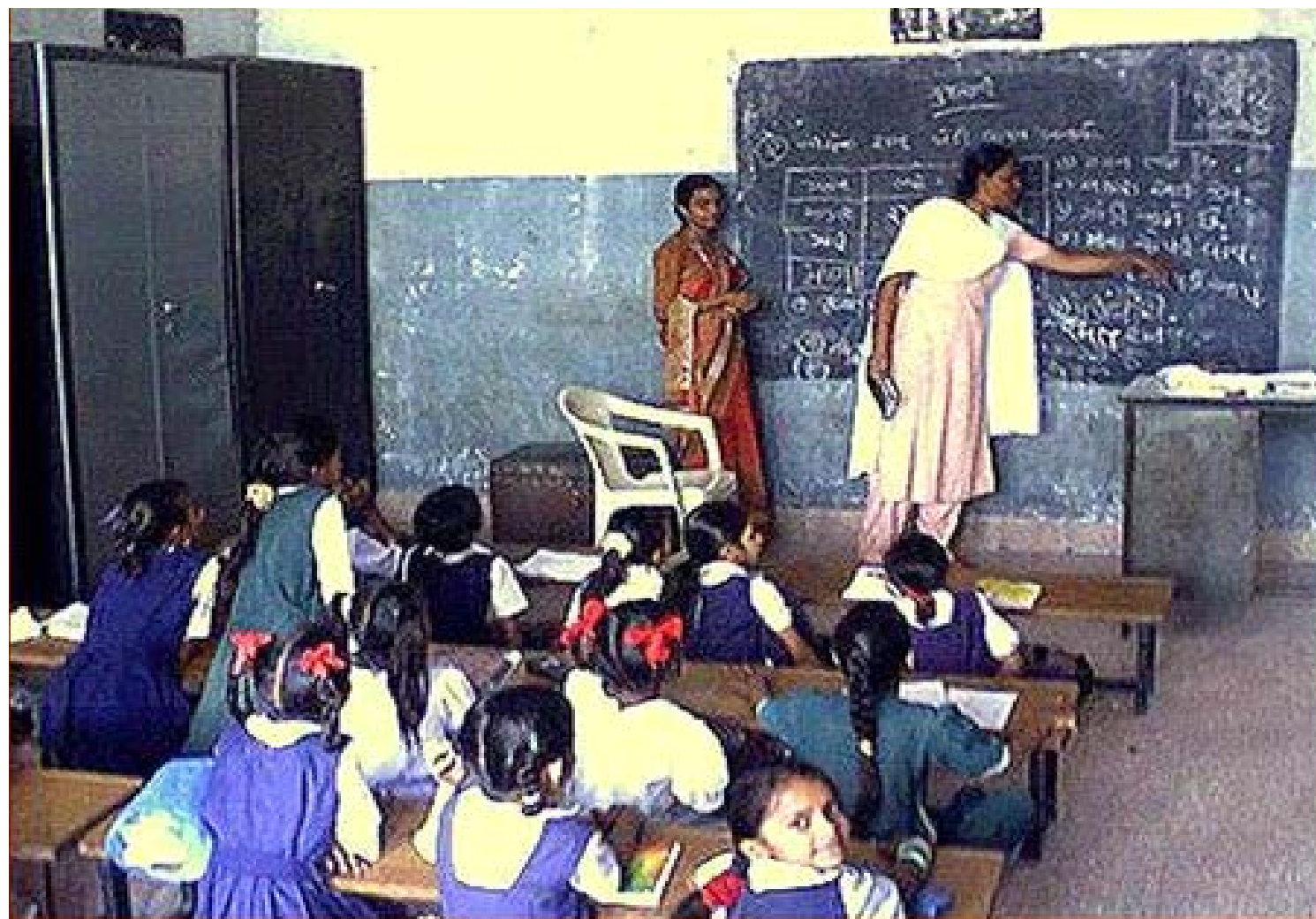
# What is the impact of this program?

- A. Positive
- B. Negative
- C. Zero
- D. I don't know
- E. Who knows?





# Example #3: Balsakhi Program



# Impact of Balsakhi - Summary

Method	Impact Estimate
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