

A decorative graphic in the top right corner consisting of numerous thin, wavy, light blue lines that flow from the top right towards the bottom right, creating a sense of movement and depth.

**TASO**

Annual  
Conference:

**How to Evaluate**

#TasoCon24

TASO

# Trials and tribulations:

Randomised controlled trials (RCTs)  
made easy

Luke Arundel / TASO

Dr Rob Summers / TASO

#TasoCon24

# The Randomistas...





# The Randomistas... Nobel Prize winners!



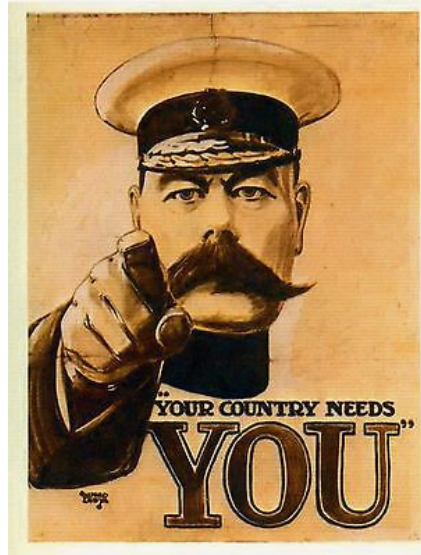
[A Nobel Prize for the Randomistas \(Jakiela, 2019\)](#)

# The Randomistas... Nobel Prize winners!

“...their **advocacy of randomised trials** is unique—few if any other methodological innovations... have so **radically altered the everyday research practices of a field** in such a short space of time.”

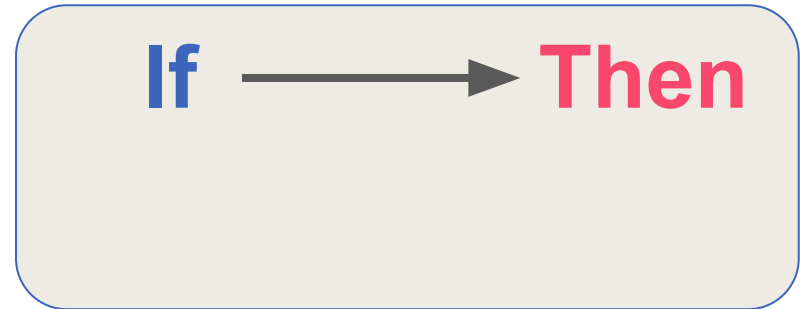
# The Randomistas of higher education

**Type 3 evidence - Causality:** research that demonstrates an activity has a 'causal impact' on outcomes for students.

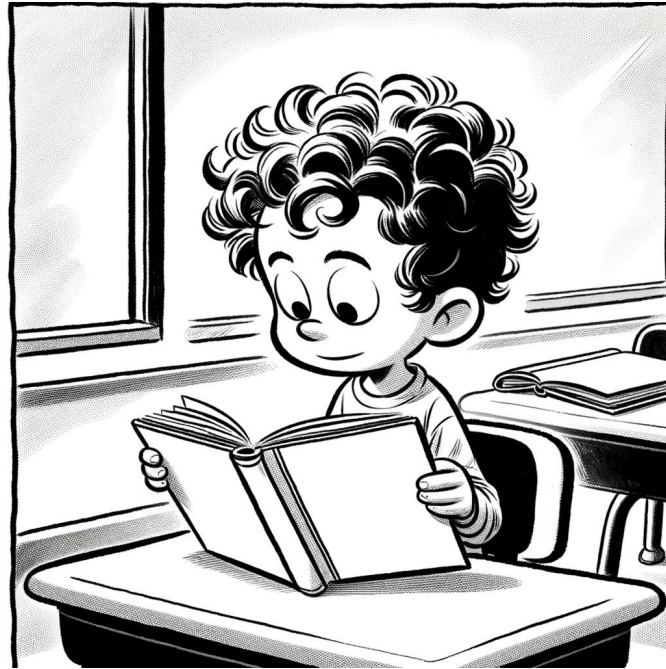


# RCTs help us get causal answers

If-then questions: if we do A,  
then will B happen?



# Do textbooks improve student outcomes?

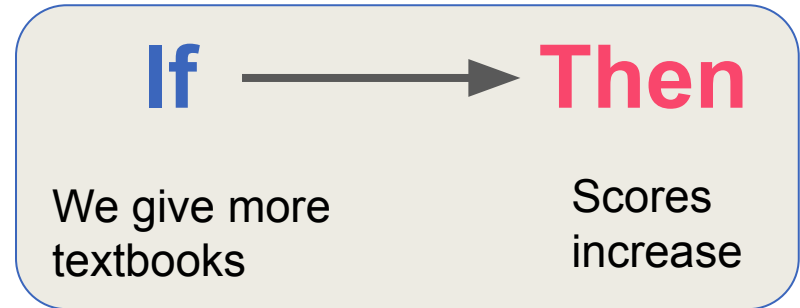




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If-then questions: if we do A,  
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If we give students more  
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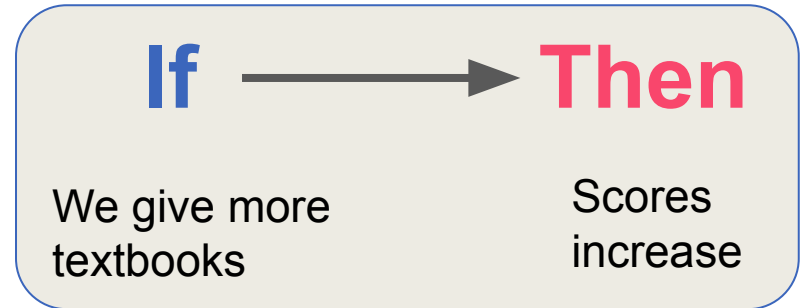


# RCTs help us get causal answers

If-then questions: if we do A,  
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If we give students more  
textbooks, then will student  
scores increase?

Intuitively - of course!



# Do textbooks improve student outcomes?

Correlation would likely support this - it works!

Note: synthetic data for chart

**Student scores increase when the number of textbooks per pupil is higher**

Textbooks per student



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Correlation would likely support this - it works!

Not so quick...

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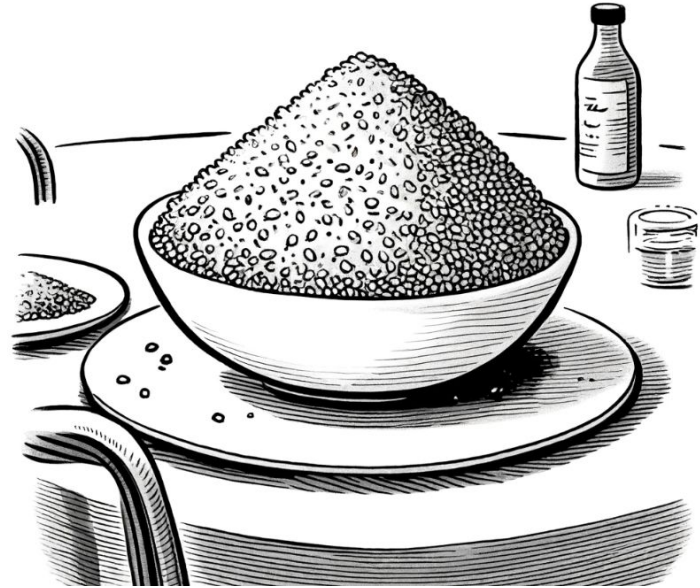
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# Correlation $\neq$ Causation: **confounders**

People who eat quinoa live longer!

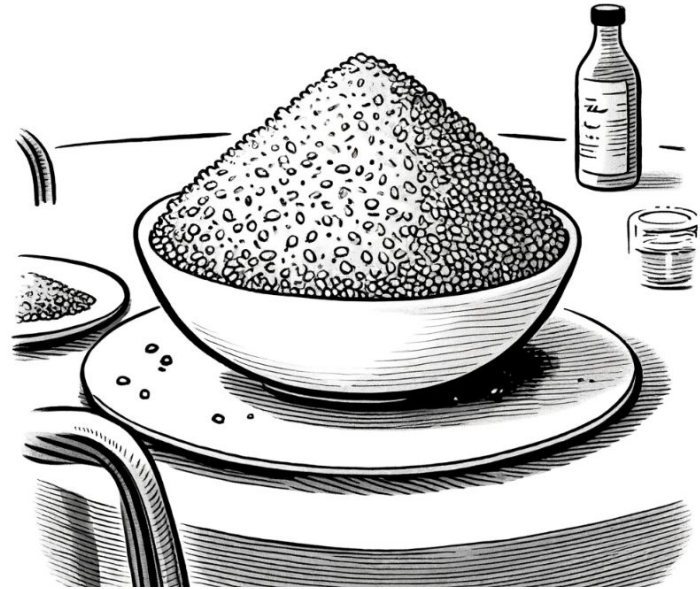




# Correlation $\neq$ Causation: confounders

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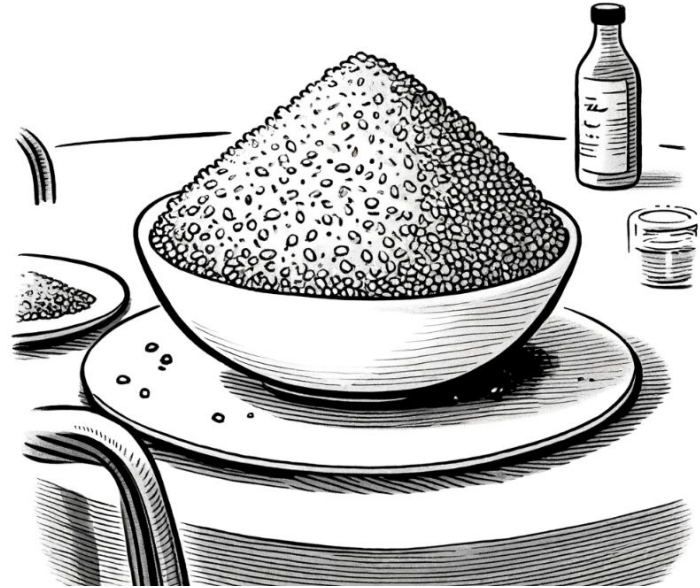


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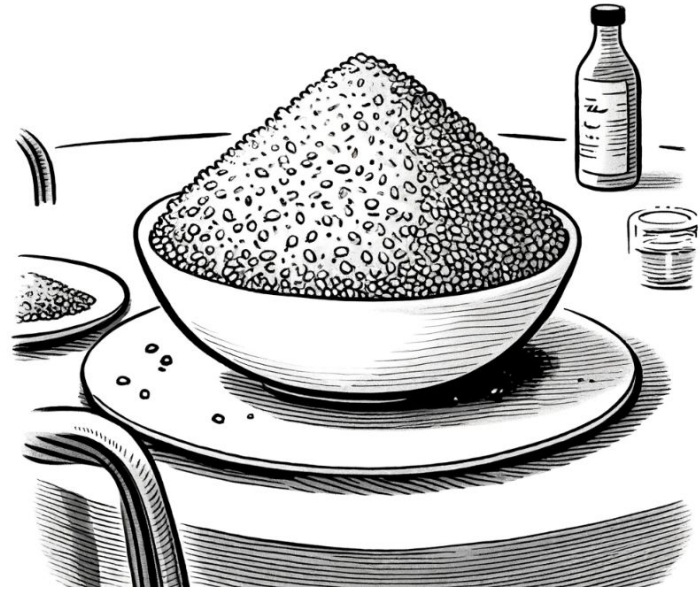
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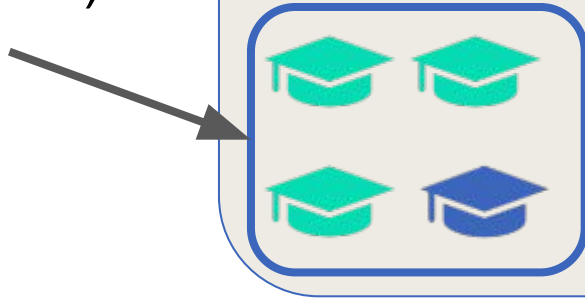
Money! → **Confounding variable**



# Correlation $\neq$ Causation: **selection bias**

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The sample  
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# Correlation $\neq$ Causation: selection bias

The sample selected  
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The whole population

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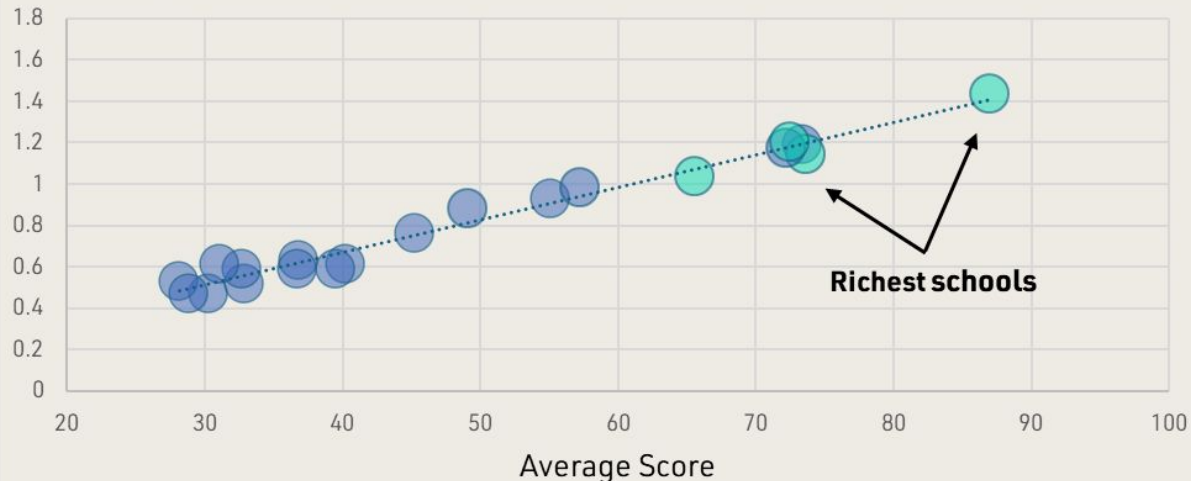
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How do we deal with this?

Note: synthetic data for chart

Student scores increase when the number of textbooks per pupil is higher

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# How do we deal with selection bias and confounding variables?



## How do we deal with selection bias and confounding variables?

If we can: **Random assignment**





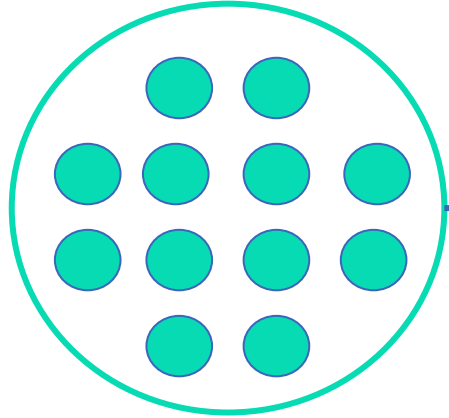
## How do we deal with selection bias and confounding variables?

If we can: **Random assignment**

If not: **Quasi-experimental designs**

“The **most credible and influential** research designs use random assignment”

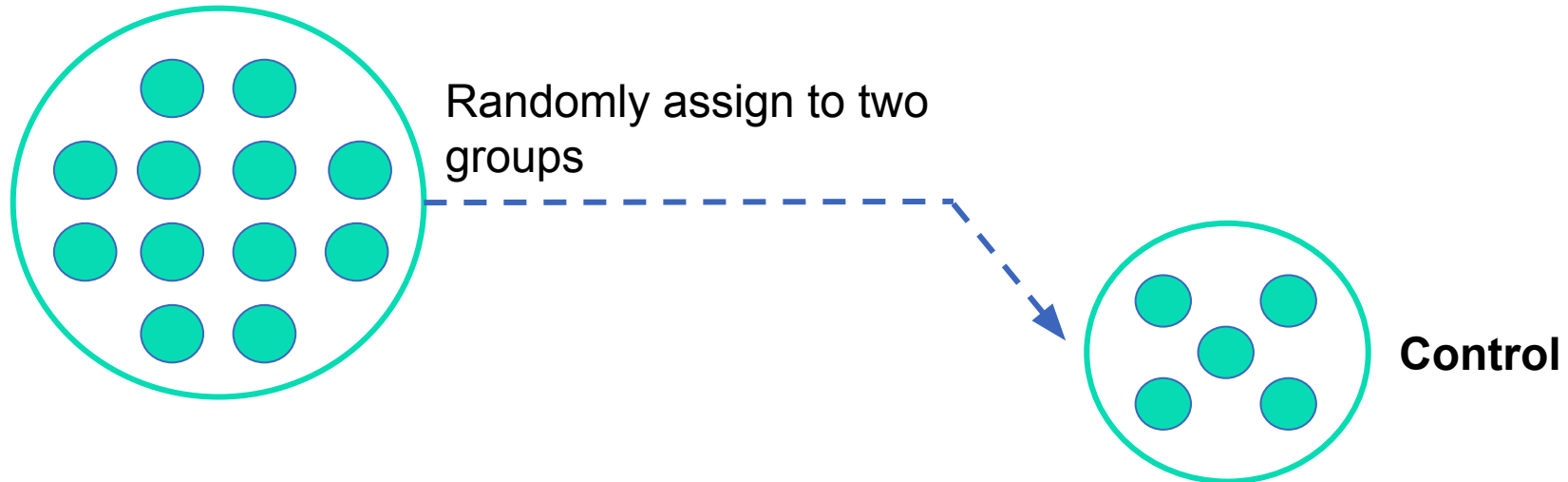
## Random assignment deals with selection bias and confounders



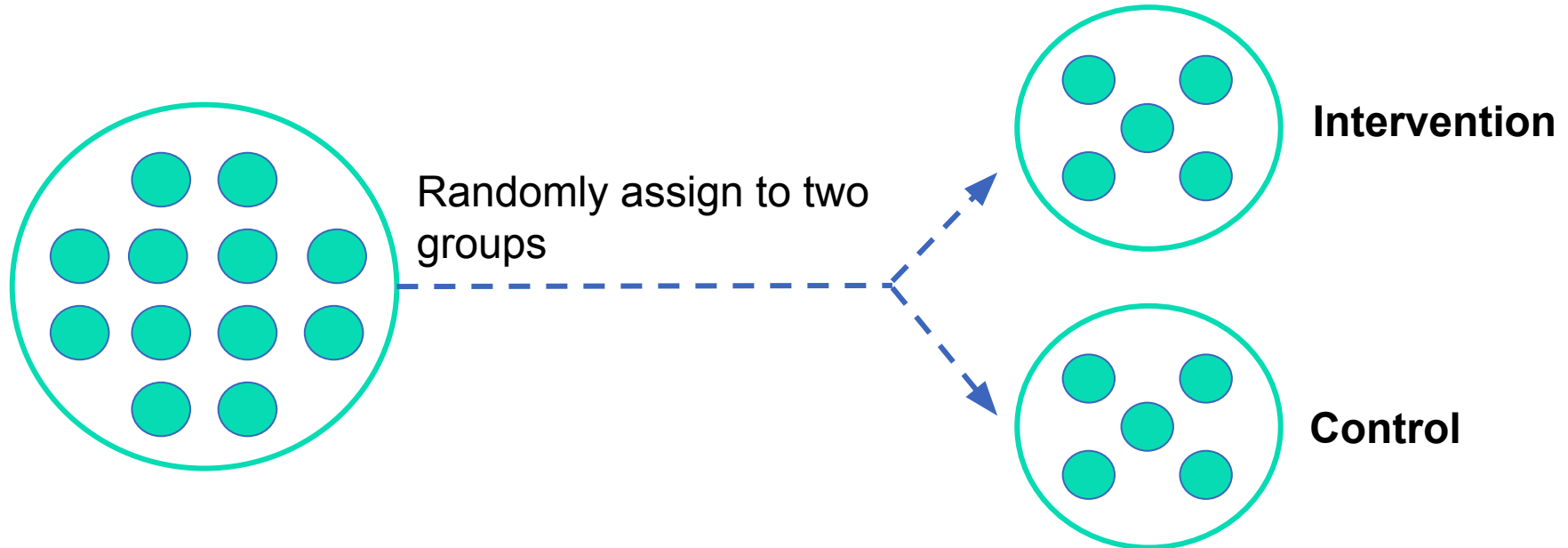
Randomly assign to two groups



## Random assignment deals with selection bias and confounders

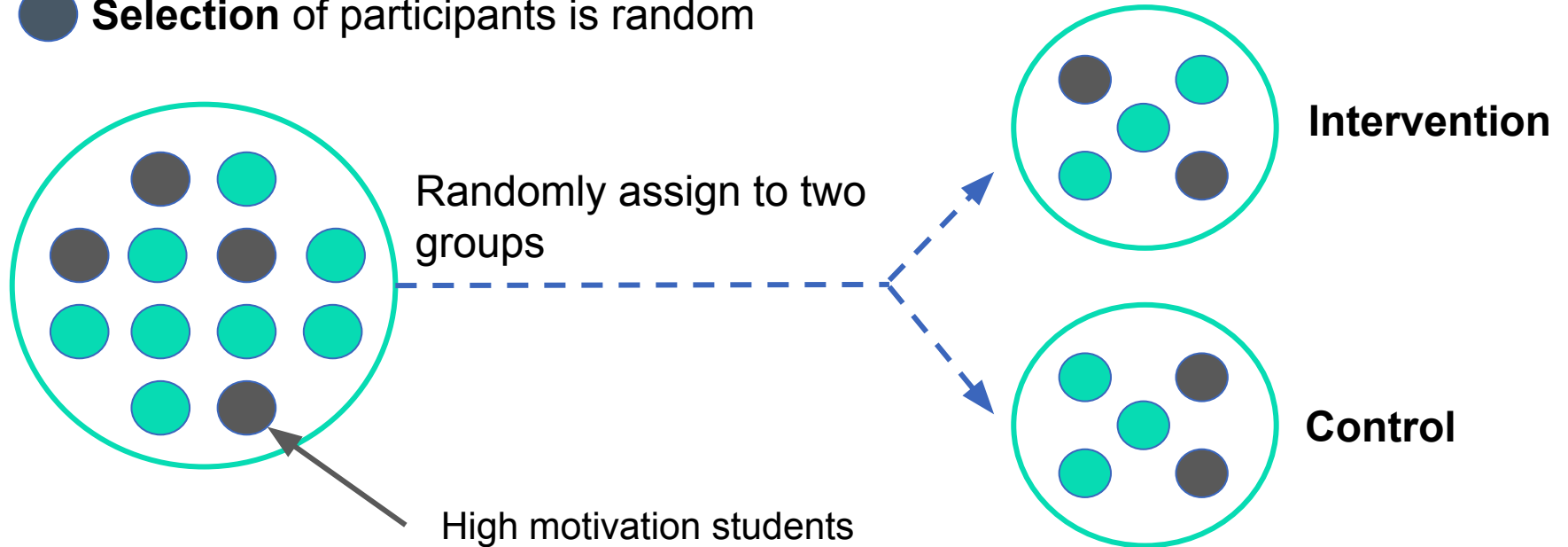


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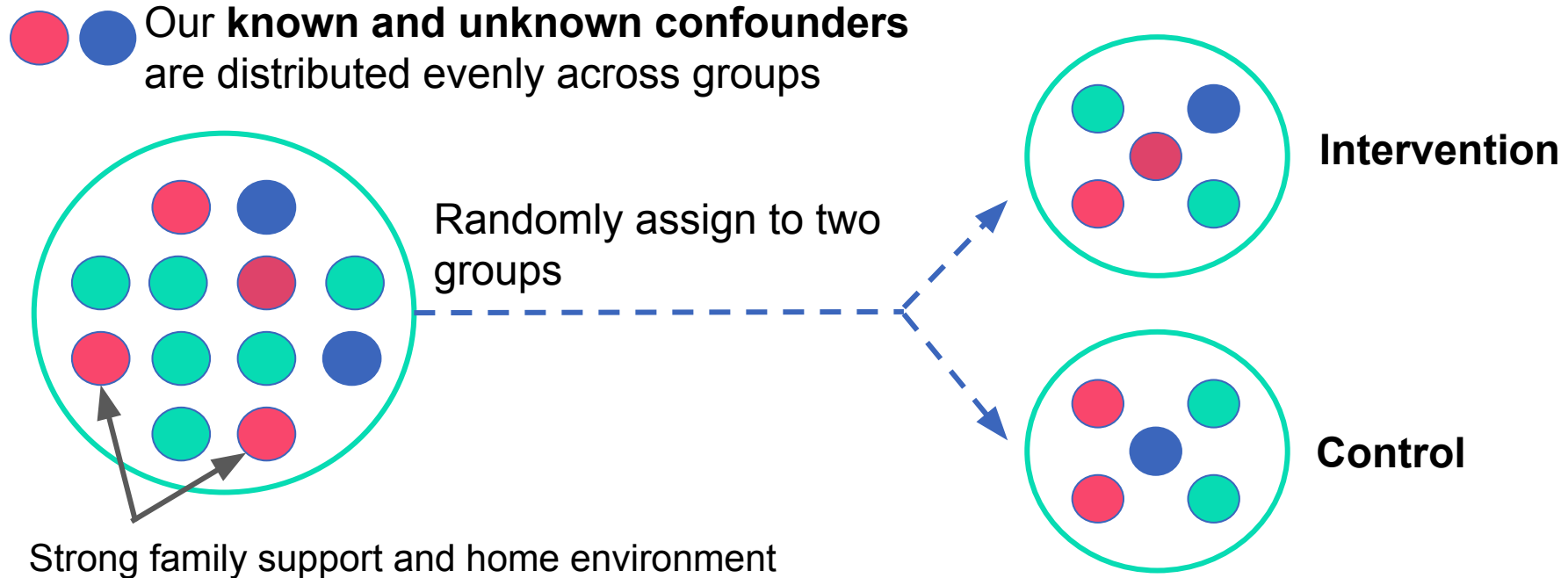


# Random assignment deals with selection bias

● Selection of participants is random

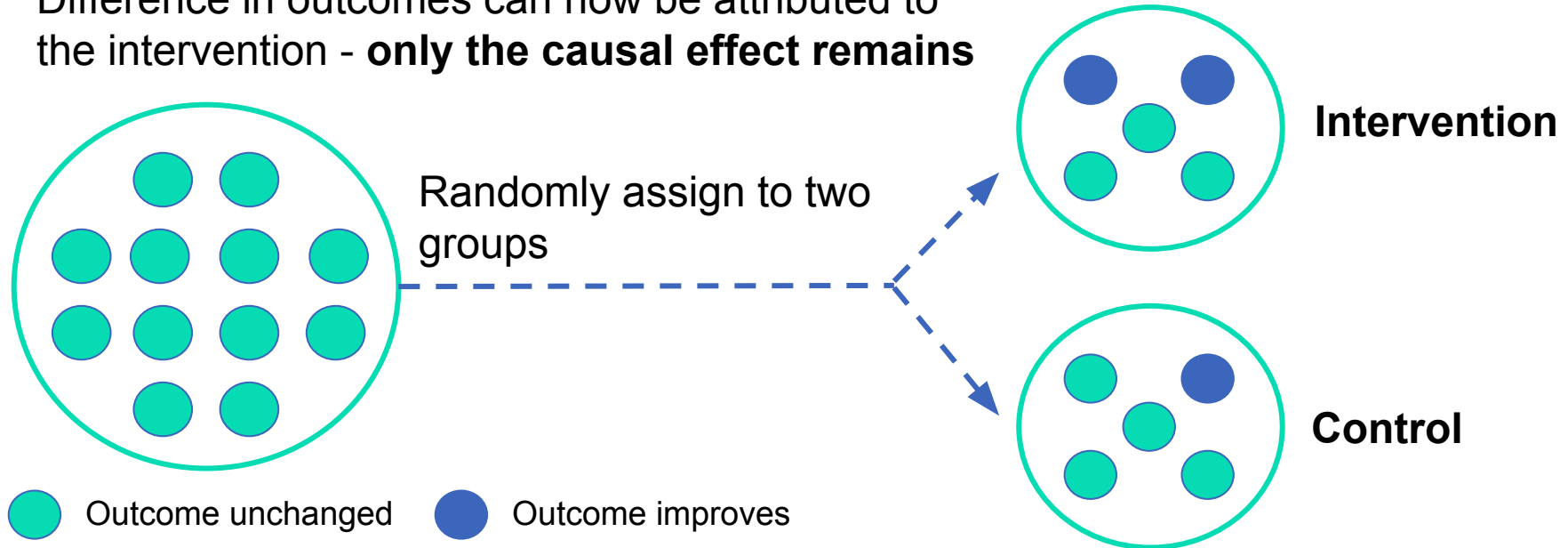


# Random assignment deals with confounders



# Random assignment deals with selection bias and confounders

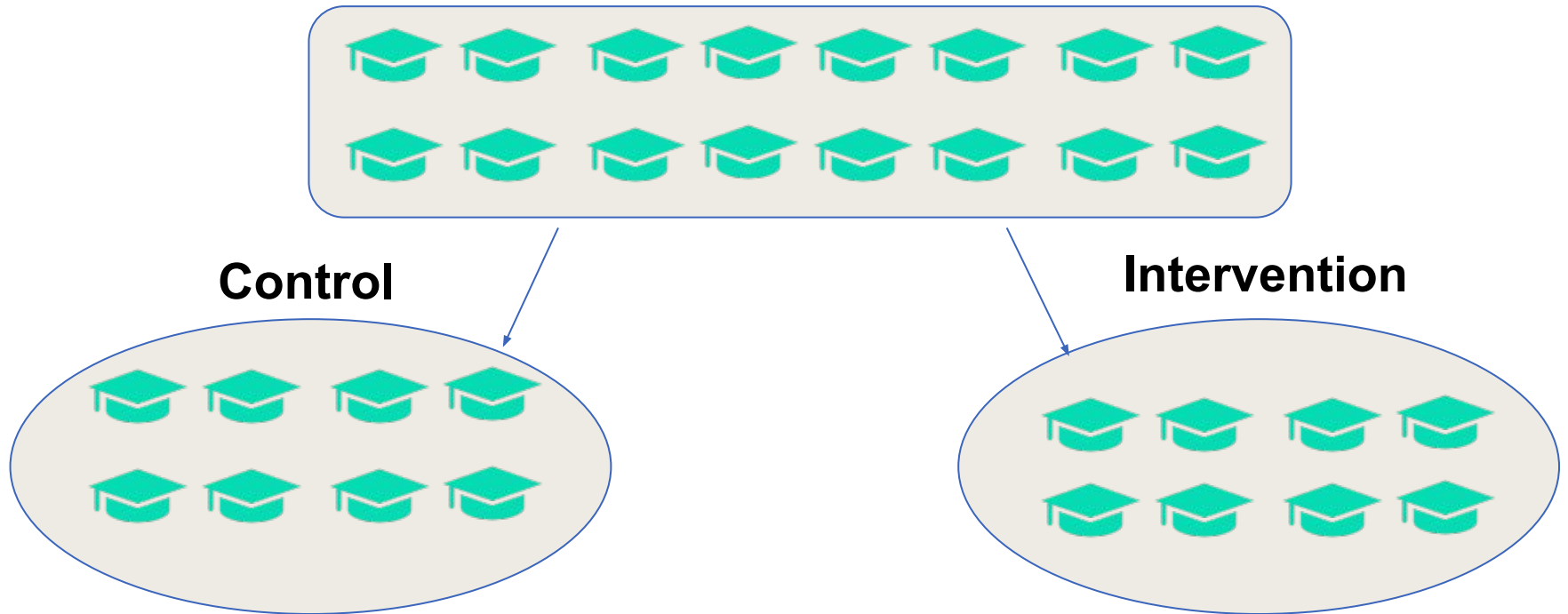
Difference in outcomes can now be attributed to the intervention - **only the causal effect remains**



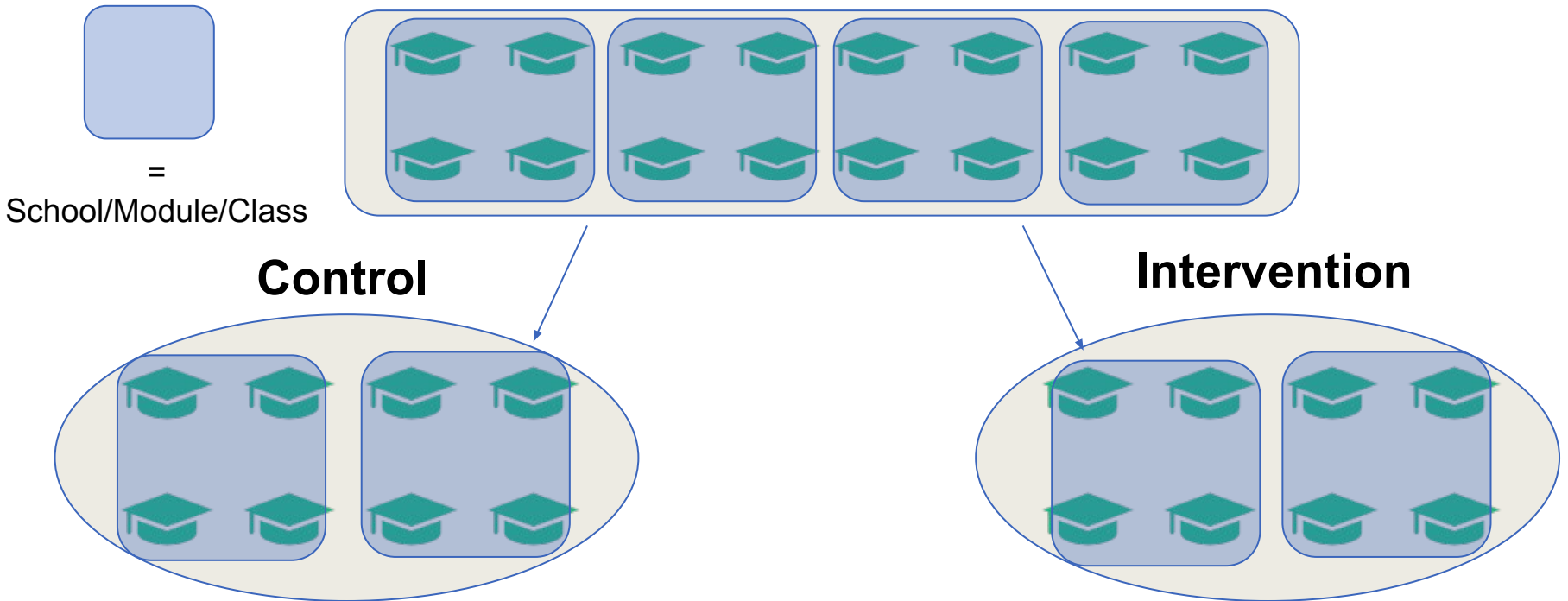
# Randomising at individual or group level



# Randomising at **individual** or group level

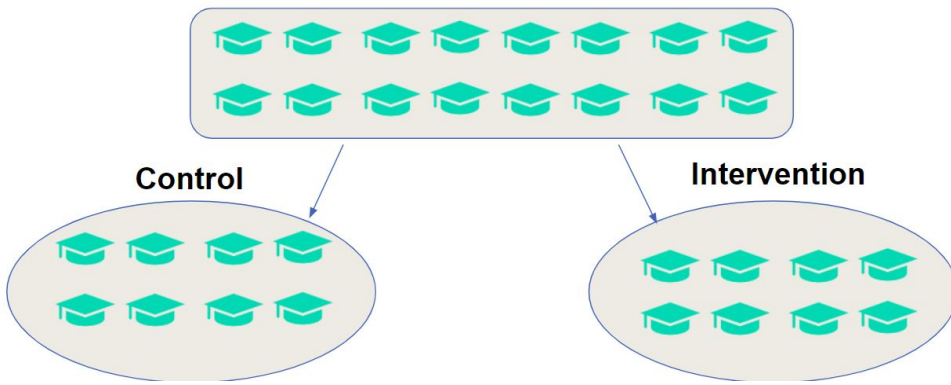


## Randomising at individual or **group** level

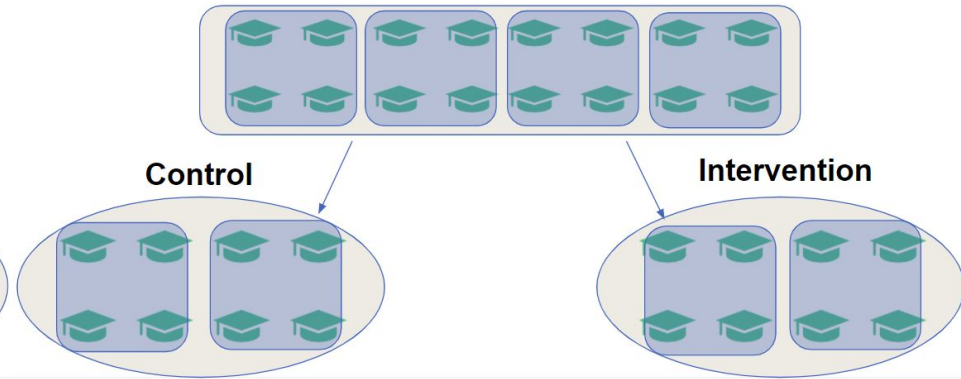


# Randomising at individual or group level

## Individual



## Cluster



# Randomising at individual or group level

<b>Individual</b>	<b>Cluster</b>
Should be favored over cluster if possible, as they are more precise and can more directly control for confounders	Less powerful than individual RCTs
Require fewer participants to detect an effect	Require more participants to detect an effect

# Randomising at individual or group level

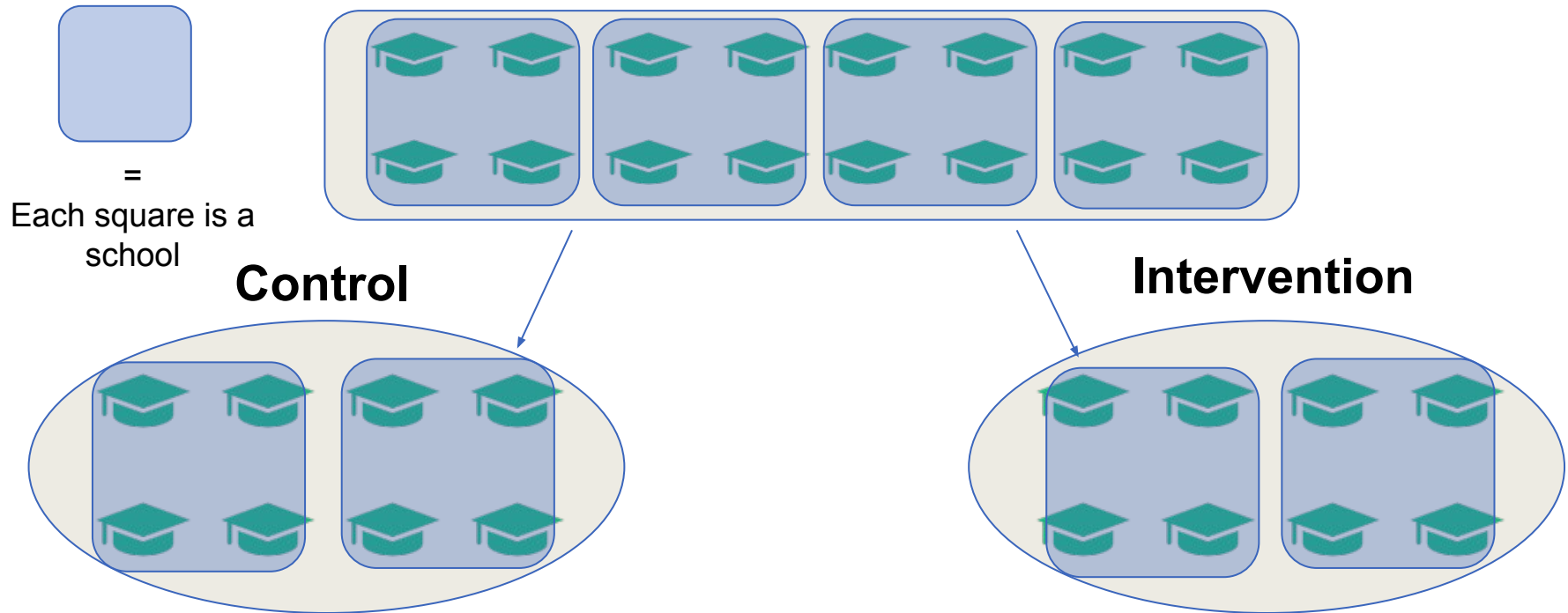
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May be impractical in some cases	Can be more practical if intervention is delivered to groups

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Require fewer participants to detect an effect	Require more participants to detect an effect
May be impractical in some cases	Can be more practical if intervention is delivered to groups
May not handle spillover effects	Better at dealing with spillover effects

Do textbooks improve student outcomes?

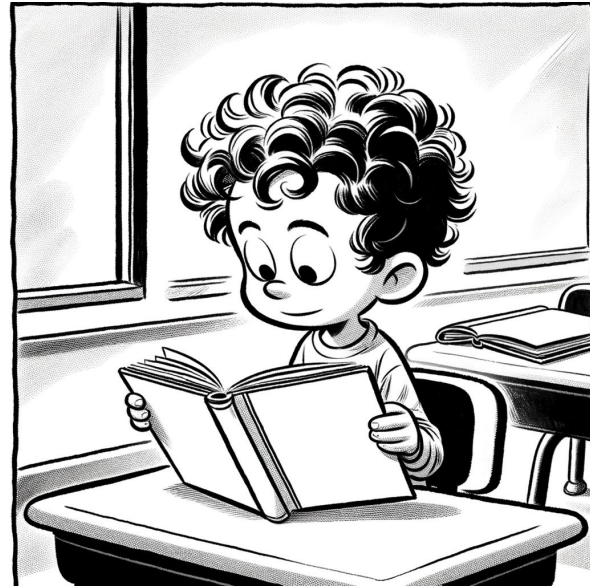
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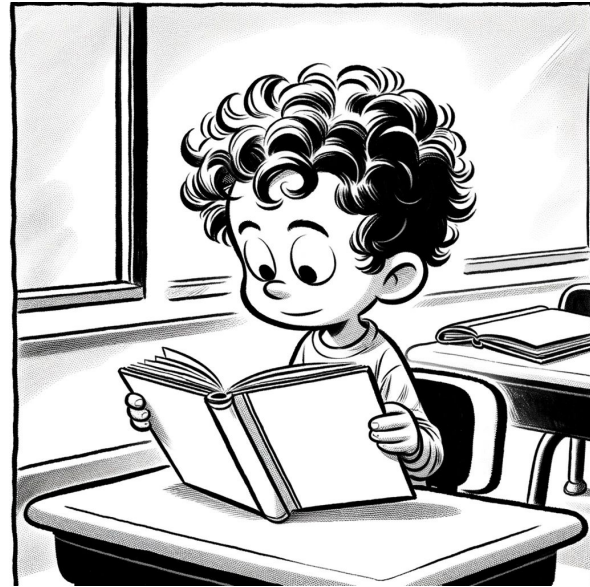
Cluster RCTs from Sierra Leone and Kenya suggest...



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Probably not (in this situation...)

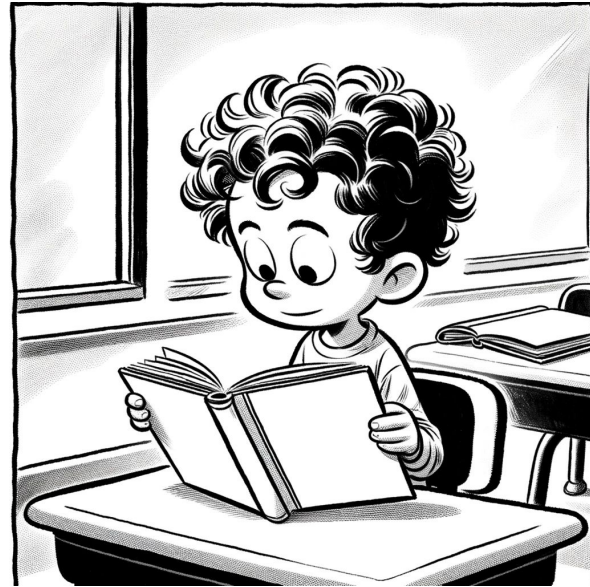


# Do textbooks improve student outcomes?

Cluster RCTs from Sierra Leone and Kenya suggest...

Probably not (in this situation...)

Implication for the huge amount spent on textbooks in Kenya



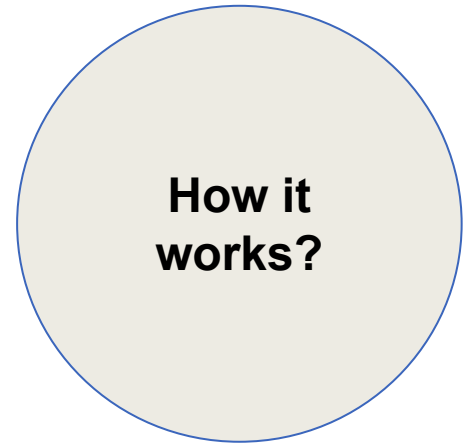
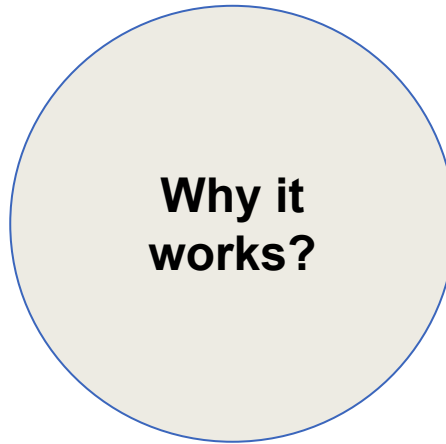
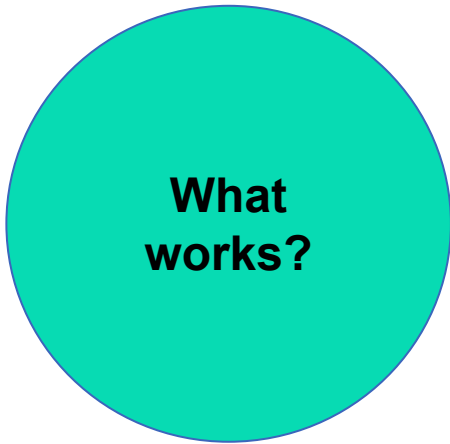
# RCTs can tell us **what** works



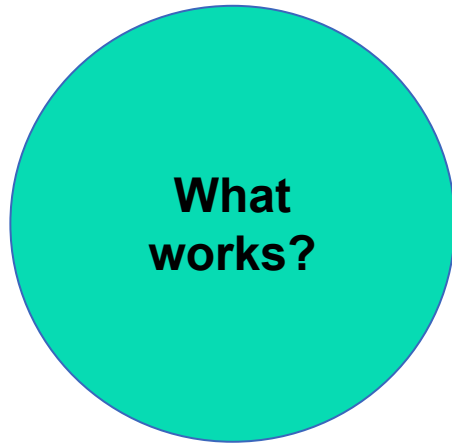
**What  
works?**



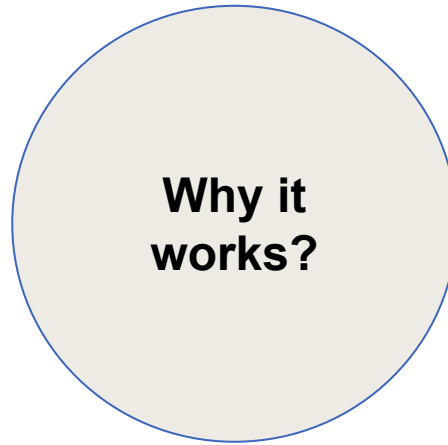
# RCTs can tell us **what** works, but not **how**



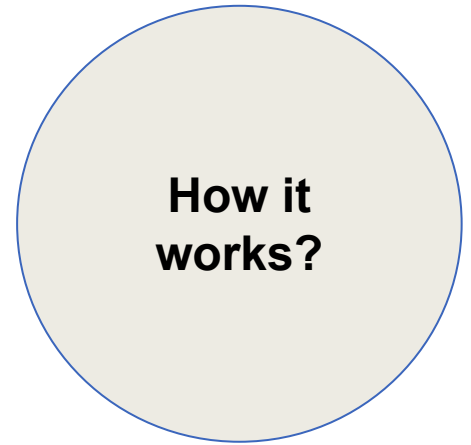
# RCTs can tell us **what** works, but not **how**



Randomised controlled trials



Theory of change and implementation and process evaluation (IPE)



# Ethics



# Ethics

- If there is substantial, consistent, high-quality evidence that something is effective, it shouldn't be withheld from anyone who might benefit
- But it is surprising how often this is not the case
- We also should think about whether the benefit justifies the cost





## Ethics: Informed consent

- Participation should be appropriately informed
- Participation in research should be voluntary
- Consent to participate can be withdrawn at (almost) any time without penalty

## Live RCT: Theory of change

### Activity

Providing participants with well-designed charts that follow principles of effective data visualisation

### Outcomes

Improved accuracy and confidence in participants' responses to information presented in the chart

### Impact

Enhanced ability of participants to interpret and use data.

### Research question

Does the use of well-designed charts improve the understanding and retention of information among participants compared to standard chart designs?

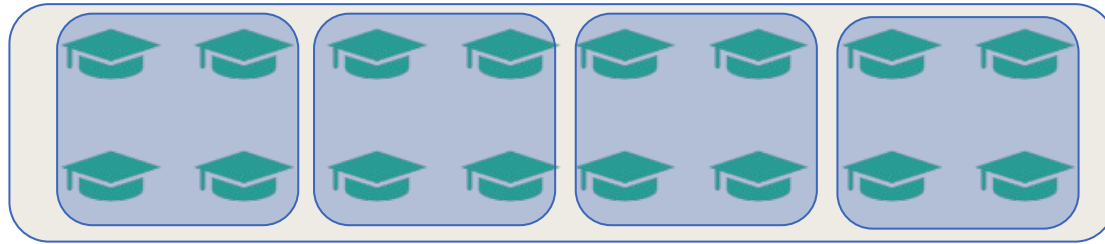
# Live RCT: randomisation unit

## Live RCT: randomisation unit

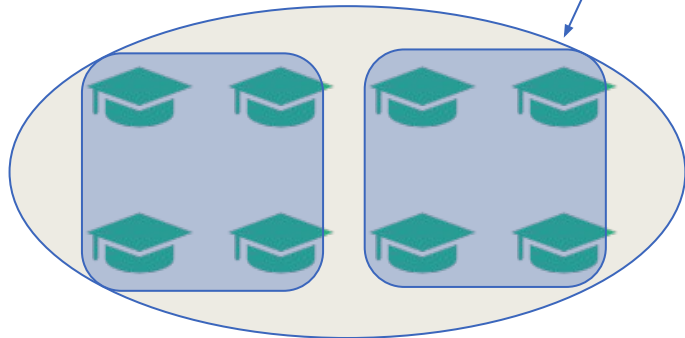


=

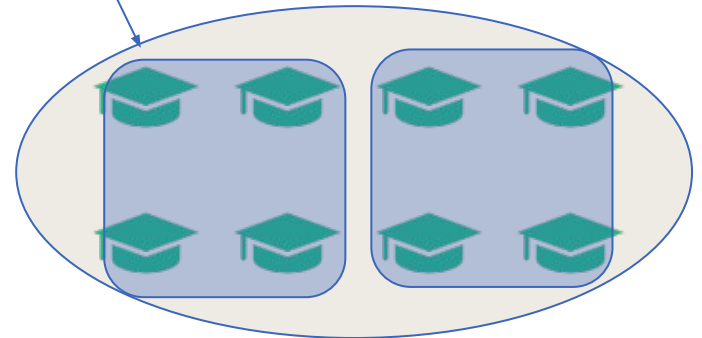
Each square is a **table**



**Control**



**Intervention**



## Live RCT: Ethics

- Participation should be appropriately informed
- Participation in research should be voluntary
- Consent to participate can be withdrawn at (almost) any time without penalty

## Live RCT

1. **Retrieve your chart:** Take one sheet each from the envelope place it chart face down on the table.
2. **View the chart:** When the timer starts turn the paper over and study the chart (you have 30 seconds)
3. **After viewing:** Once the 30 seconds are over, turn the paper over **do not look at the chart again.**
4. **Answer the questions:** Please use the QR code or weblink to access the questions.

## Live RCT

1. **Retrieve your chart but don't look at it yet**
2. **View the chart** when the timer starts (you have 30 seconds)
3. Once the 30 seconds are over, turn the paper over **do not look at the chart again.**
4. **Answer the questions:** Please use the QR code or weblink to access the questions.



# Live RCT: results



## Live RCT: results

### Research question

Does the use of well-designed charts improve the understanding and retention of information among participants compared to standard chart designs?

Is there the predicted effect?

Any backfire effects?

Spillover?

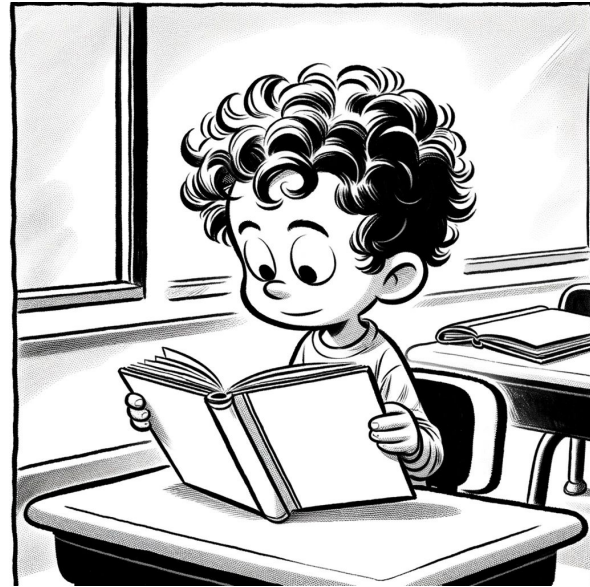
Differences in demographics?

Importance of the relationship between RQ and intended effect/outcome

# Do textbooks improve student outcomes? : Implementation and Process Evaluation (IPE)

Trials suggested providing  
textbooks didn't work

IPE revealed why...



## Do textbooks improve student outcomes? : Implementation and Process Evaluation (IPE)

Trials suggested providing textbooks didn't work

IPE revealed why...

Not aimed at the right level and not distributed



# Live RCT: implementation and process evaluation

What drove the impact?

Fidelity of the trial

What other data would we want to collect?



## Other considerations...

# Other considerations... sample size

Need a large enough sample size to make confident conclusions

Clusters will need more

Depends on the size of the anticipated effect

[Test, Learn, Adapt \(Haynes et al. 2012\)](#)

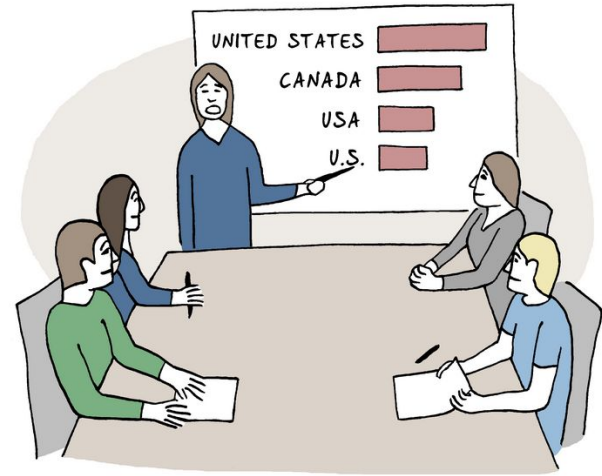
<u>P-VALUE</u>	<u>INTERPRETATION</u>
0.001	HIGHLY SIGNIFICANT
0.01	
0.02	
0.03	
0.04	SIGNIFICANT
0.049	
0.050	OH CRAP. REDO CALCULATIONS.
0.051	ON THE EDGE OF SIGNIFICANCE
0.06	
0.07	HIGHLY SUGGESTIVE, SIGNIFICANT AT THE P<0.10 LEVEL
0.08	
0.09	
0.099	HEY, LOOK AT THIS INTERESTING SUBGROUP ANALYSIS
≥0.1	

## Other considerations... data availability

You need outcome data for both participants and non participants

Demographic and contextual data for intervention participants and non-participants

See Data Infrastructure Guide this summer for more



AS YOU CAN SEE, OUR TOP MARKETS ARE UNITED STATES, CANADA, USA AND THE U.S.

## Other considerations... **resourcing**

Randomisation, data collection, data recording, and data-quality checking

Design, pre-register and run the analyses

Implementation and process evaluation

Reporting



## Other considerations... cost

With planning, can be cheaper than you would expect

Particularly where service is already being delivered and data already being collected

What are the costs of not doing an RCT?

[Test, Learn, Adapt \(Haynes et al. 2012\)](#)



## Other considerations... **cost-benefit**

Intervention probably costs a lot of money - is the investment worth it?

Can do proper cost-benefit using causal impact - See upcoming TASO Economic Evaluation Framework

We have limited resources so I'm going to suggest we only fund projects that work really well.



freshspectrum.com

## Conclusion

RCTs help us answer causal questions

They do this by eliminating the issue of confounders and selection bias

QEDs can be more appropriate, but well-designed RCTs are the most robust evaluations

IPE complements RCTs: RCTs tell you what works, IPE tells you how/why

Feasibility of an RCT depends on many factors, such as required sample size, data availability, resourcing

# Conclusion



[The Experimental Approach to Development Economics \(Banerjee & Duflo, 2009\)](#)

# Conclusion

It's important to know what does and doesn't work, and for this we need causal evidence

This can be done through quasi-experimental designs, but the 'experimental ideal' is an RCT



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Q&A

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RCT webinar -  
sign up now!



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# Lunch break

13:00–14:00

Next: Breakout sessions: Unlocking the  
evaluation toolkit

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