

# Quantitative Research Methods

# Quasi-Experimental Design



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# Quasi-Experimental design differs from true experimental design in its non-random group assignment

## "True" experimental designs



**Random**  
assignment

- ✓ Gold standard for quantitative research
- ! Truly random assignment can be challenging or ethically concerning



## Quasi-experimental designs



**Non-random**  
assignment

- ✓ Often easier to implement
- ! Challenges with internal validity

Focus of presentation

# Today's agenda



Nonequivalent Groups Design



Regression-Discontinuity Design



Other Quasi-Experimental Designs

# Today's agenda



Nonequivalent Groups Design



Regression-Discontinuity Design

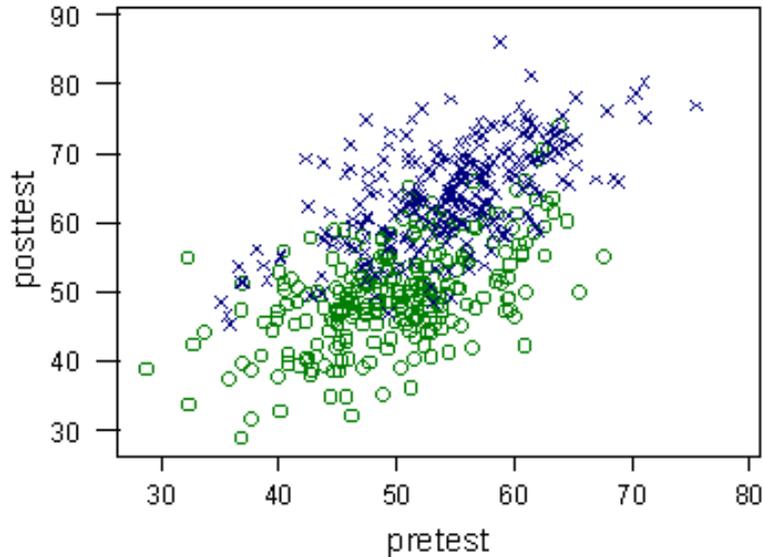


Other Quasi-Experimental Designs

# Nonequivalent groups design uses already existing groups for treatment and control

- Non-equivalent groups design (NEGD) is structured like **pretest-posttest randomized experiment without random group assignment**
- **Treatment and control group are formed based on previously existing groups**, i.e., in a school using two comparable classrooms or using two similar communities
- Even if similar groups are selected, they are unlikely as similar as if chosen at random -> therefore "**nonequivalent groups**"
- Prior differences in groups raises **internal validity threat of selection**, meaning that differences of groups may affect outcome of study

# Bivariate distribution reveals issue of selection threat to internal validity



 Control cases

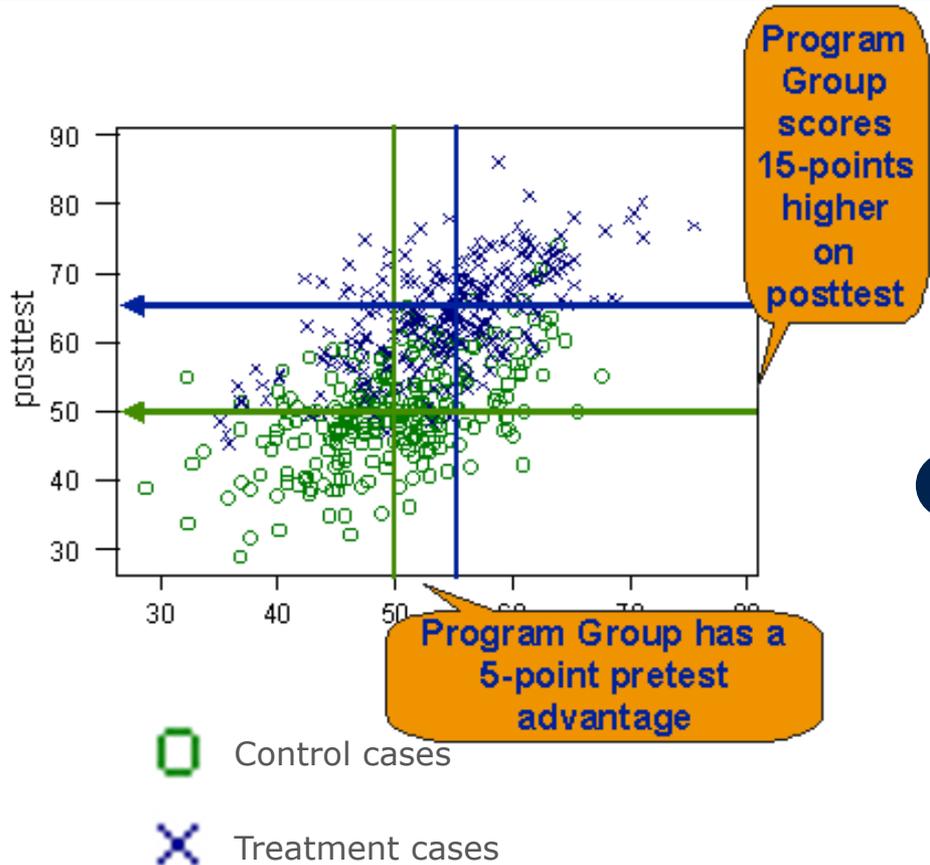
 Treatment cases

Looking at hypothetical bivariate distribution of NEGD reveals two insights:

1. **Treatment cases clearly score higher on posttest** than control cases (y-axis)
2. On average, **treatment cases performed slightly better on pretest already** (x-axis)

 Hard to determine, whether difference is partially or fully caused by treatment or if initial advantage led to better outcomes (selection threat)

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- Hard to determine, whether difference is partially or fully caused by treatment or if initial advantage led to better outcomes (selection threat)

# Most important types of selection threats to internal validity

## Selection-maturation threat

Implies that groups are maturing at different rates, which creates illusion of program effect

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## Selection-history threat

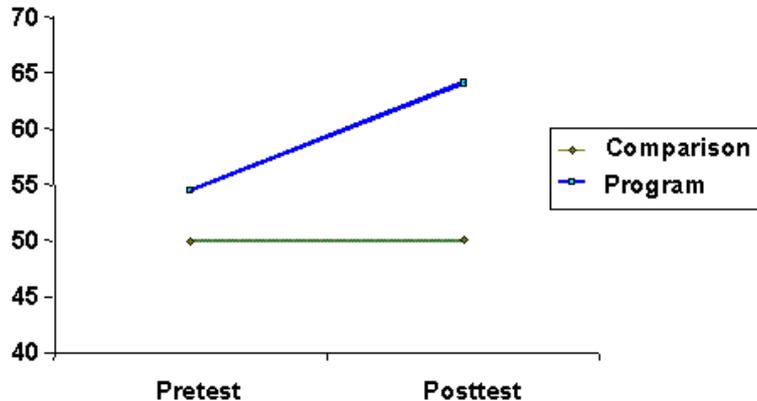
Occurrence of event that only affected one group or that only occurred for one group

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## Selection-regression

Implies that observed differences stem from sample mean of group regressing to population mean over time

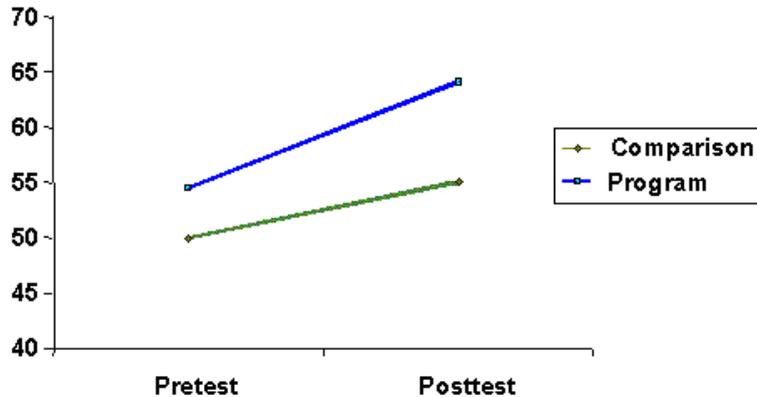
# Exemplary graph #1: No clear evidence for effectiveness can be inferred



Different representation of previous graph – what possible threats apply?

- **Selection-maturation threat**
  - Unlikely, as no maturation at all can be observed in comparison group
- **Selection-history threat**
  - Plausible for case due to differences in development
- **Selection-regression**
  - Unlikely, as upwards trend in program group implies they were below population mean – if regression to the mean was cause we would see it in comparison group too

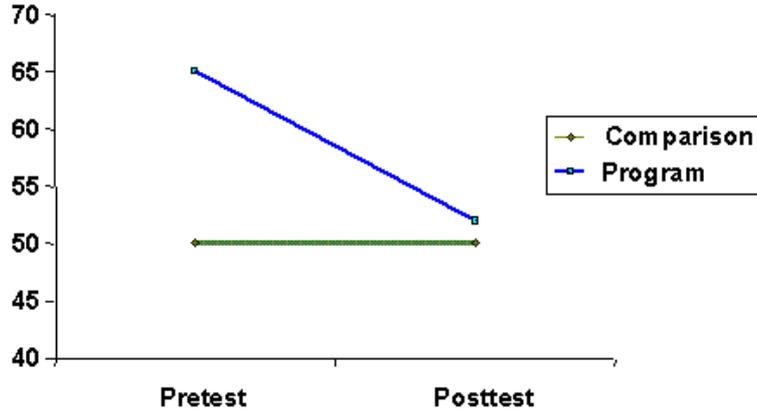
# Exemplary graph #2: No clear evidence for effectiveness can be inferred



What possible threats apply?

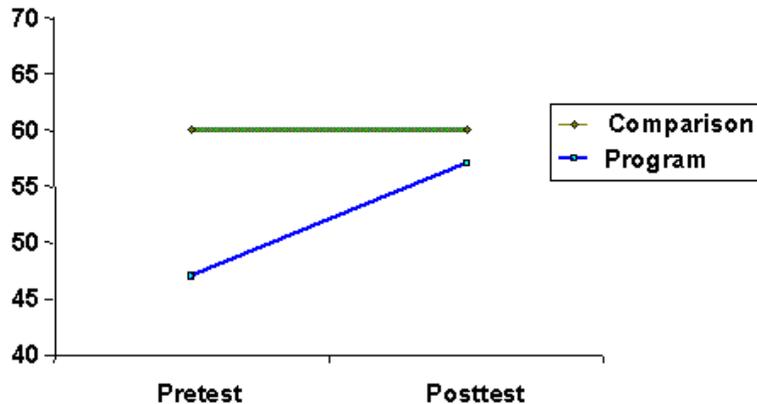
- **Selection-maturation threat**
  - Likely, as previous differences might have arisen from different maturation rates, which further increased
- **Selection-history threat**
  - Also likely, if groups react differently to some event, which causes different developments
- **Selection-regression**
  - Unlikely, as upwards trend in program group implies they were below population mean – if regression to the mean was cause we would see it in comparison even more strongly

# Exemplary graph #3 & #4: No clear evidence for effectiveness can be inferred

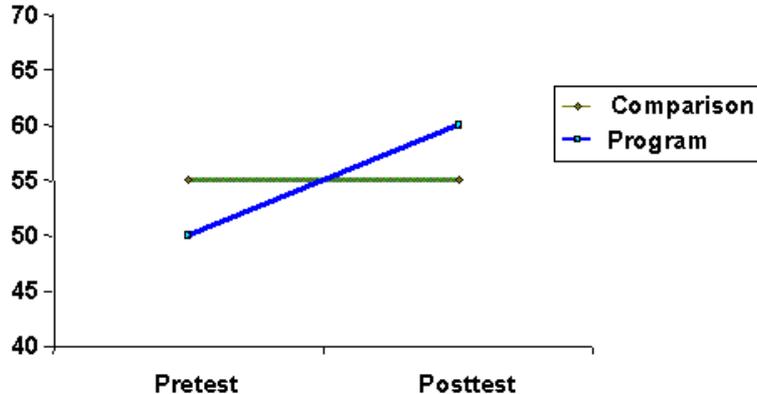


What possible threats apply?

- **Selection-maturation threat**
  - Unlikely, as maturation in program group with no maturation in comparison group seems hard to explain
- **Selection-regression**
  - Very likely -> program group was very high/low on pre-test and simply regressing to population mean, whereas comparison group was already at population mean levels



# Exemplary graph #5: Cross-over pattern implies strongest evidence for effectiveness



What possible threats apply?

- **Selection-maturation threat**
  - Unlikely, as one would need to argue that program group matured from below average to above average with no maturation in comparison group
- **Selection-history threat**
  - Unlikely, as program group started out worse off and improved beyond comp
- **Selection-regression**
  - Unlikely, as program group would need to approach comparison group, but not cross over, while no movement in comparison group

**Cross-over pattern implies strong evidence for treatment effect**, however one shouldn't structure research intentionally in such way (treating disadvantaged group and hoping it even outperforms comparison group)

# Today's agenda



Nonequivalent Groups Design



Regression-Discontinuity Design



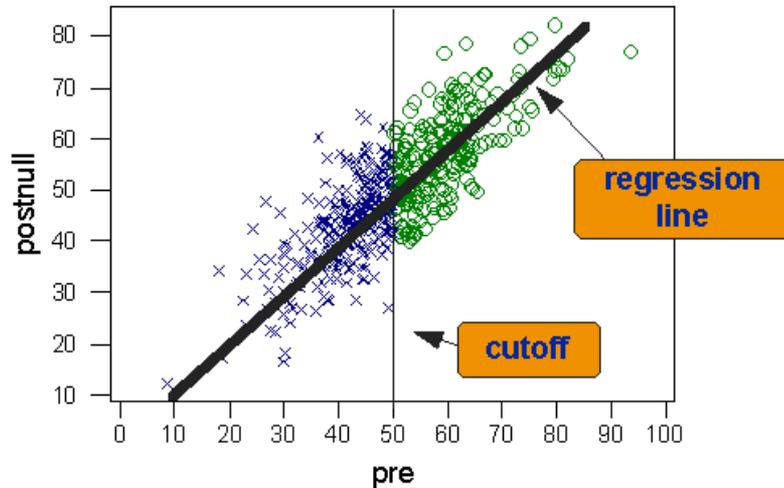
Other Quasi-Experimental Designs

# Regression-Discontinuity Design (RD)

- **Regression-Discontinuity Design (RD design)** can refer to several design variations - simplest traditional form **represents pretest-posttest program-comparison group strategy**
- **Assignment to groups is based on a cutoff score** on a pre-program measure
- Biggest advantage is that it allows to **assign treatment to those who need it most**
- **Internal validity comparable to randomized experiments**, however statistical power lower by a factor of  $\sim 2.75^1$
- **RD design not yet frequently implemented**, as its rather novel (first introduced in mid 1970s), not flexible (single quantitative measure determines group assignment), and **counterintuitive as it maximizes group differences** instead of trying to have similar groups

1. E.g., 100 participants required for significance in randomized experiment, then 275 needed for RD

# Bivariate distribution without any intervention shows continuous distribution

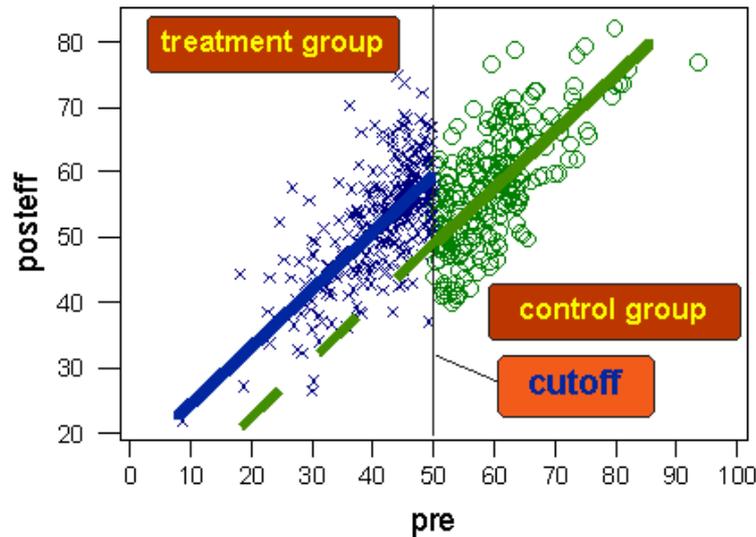


○ Control cases

× Treatment cases

- Example of cases where composite health score (100= healthy, 0 = not healthy) is used as pre-test and post-test
- Chart shows bivariate distribution of cases **without any program intervention**
- On average, being healthy on pretest means being healthy on posttest
- Smooth regression line can be put on data points

# Bivariate distribution with intervention shows discontinuity in regression line

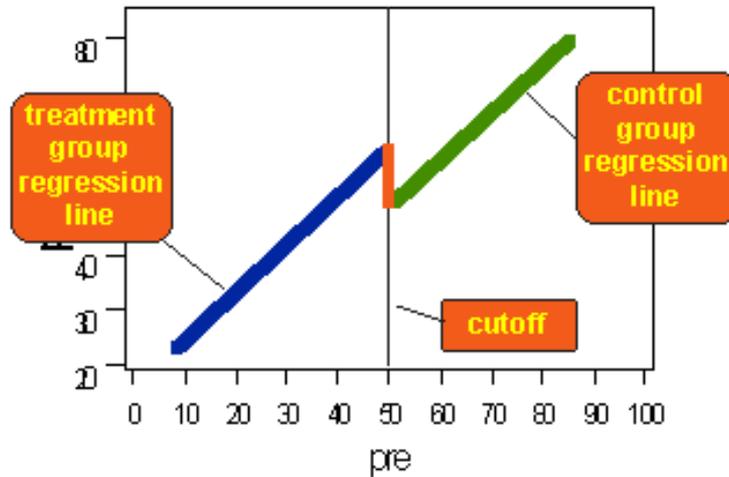


○ Control cases

× Treatment cases

- Hypothetical distribution, assuming that treatment has a constant positive effect
- Green dotted line implies expected regression line if treatment had no effect
- Under presence of a treatment effect, there will be disruption (or discontinuation) in regression line -> therefore Regression Discontinuity Design

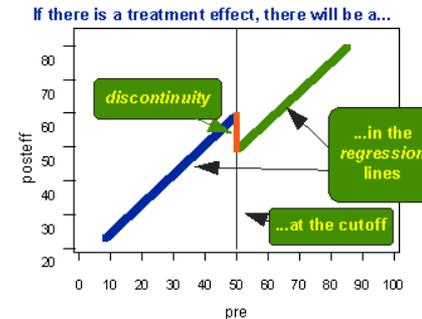
# Bivariate distribution with intervention shows discontinuity in regression line



- Control cases
- ✕ Treatment cases



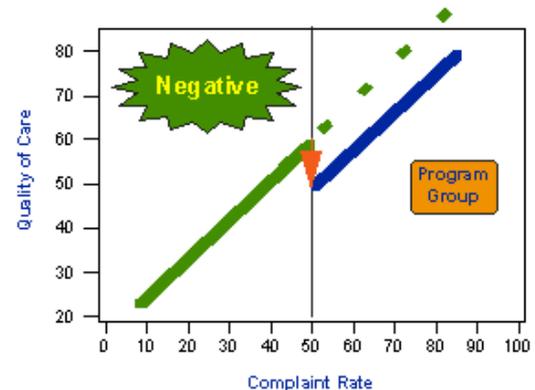
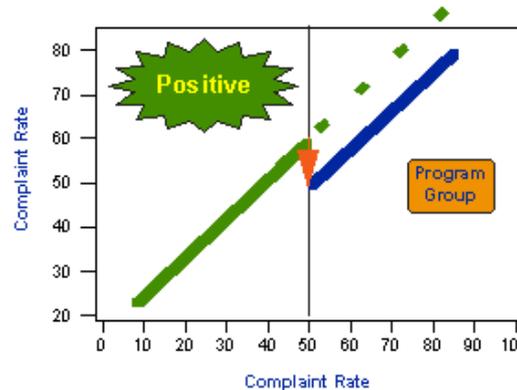
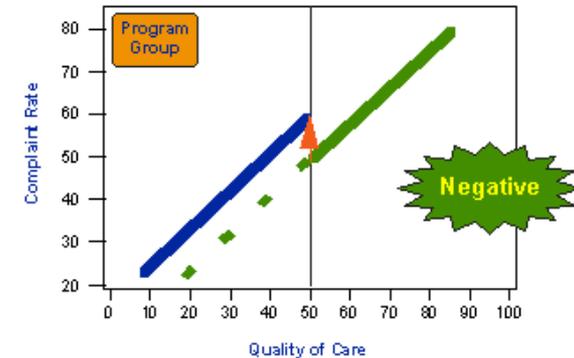
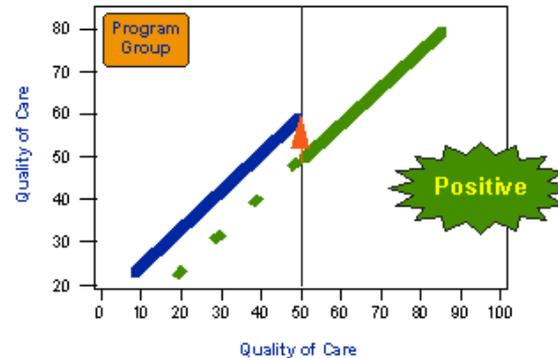
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# Interpretation of results depends on nature of assignment and outcome variable

Example:

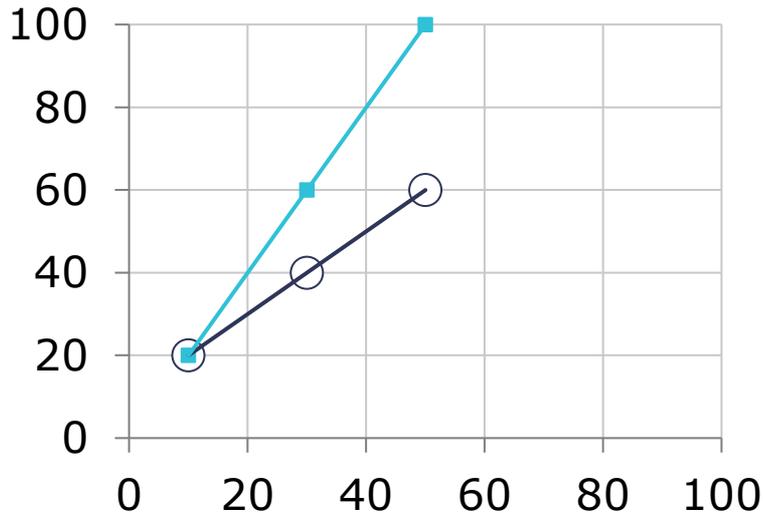
- Hospital staff to receive treatment to improve quality of care
- Two measures available: Quality of care (judged by supervisor) and complaint rate (number of complaints relative to patients)



# RD design requires a continuous pre-program variable and a defined cutoff value

- RD design requires a continuous quantitative pre-program measure (which can but doesn't have to be different from the pretest measure)
- All persons on one side of cutoff are assigned to treatment, all persons on other side are assigned to control
- Selection of cutoff
  - **Based on project resources:** I.e., if 25 people can be treated, cutoff can be set such that 25 fall into treatment and rest into control
  - **Based on expert views:** I.e., if expert believe health score of 50 or lower indicates treatment, then 50 can be set as cutoff
- In other designs, we assume or provide evidence that treatment and control group are equivalent and that differences can be attributed to program
- In RD design we instead assume that in absence of program pre-post-relationship is equivalent for both groups – this assumes the following:
  - No spurious discontinuity that coincides with cutoff point
  - Correct modelling of the pre-post-relationship
  - **Introduces possibility of selection threat to internal validity**

# Selection maturation threat to internal validity

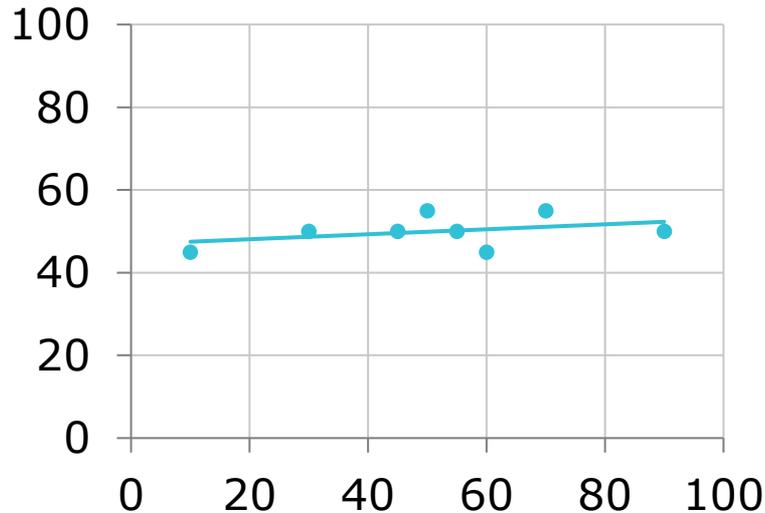


- A: Absolute maturing rate
- B: Relative maturing rate



- Selection maturity threat implies that different groups could have different rates of maturing
- I.e., group A shows a constant absolute maturing rate of +10 from pretest to posttest
- I.e., group B shows constant relative maturing rate of x2, which leads to differing absolute changes from pre to post, depending on the pre-level
- Both scenarios can be captured by a regression line, so long that differences in maturing rates do not coincide with the cutoff point

# Selection regression threat to internal validity



- Regression to the mean would imply that sampled cases approach the population mean over time
- The expected regression to the mean from pretest to posttest score is exactly what is modelled by the regression line itself
- As we do not expect a difference in regression to the mean between any two groups, there should not be a discontinuity in the regression line without intervention

# RD design shows strong internal validity with some practical caveats to be considered

- RD design by itself is not susceptible to selection threats and shows in theory the same internal validity as a randomized experiment



There is some caveats that can practically hinder internal validity:

- **Participants can manipulate their pre-program measure**, making it not random – i.e., if using students' grades as pre-program measure and 50% as the cutoff, teachers might be giving students slightly below 50% a mercy pass lifting them above 50% and thus changing their group assignment
- **There is another treatment coinciding with the cutoff of the actual treatment** – i.e., studying alcohol's effects on mental health using legal drinking age as cutoff coincides with legal gambling age, potentially contaminating the results
- **Correct modelling of the pretest-posttest relationship** – incorrect modelling of non-linear relationships could be mistaken as discontinuity

# Today's agenda



Nonequivalent Groups Design



Regression-Discontinuity Design



Other Quasi-Experimental Designs

# Proxy Pretest Design

- In general same setup as normal pretest-posttest design
- However, no pretest conducted, instead proxy measure used instead of pretest:
  - **Recollection proxy pretest:** Participants own assessment, what they believe their pretest score might have been
  - **Archived proxy pretest:** Using measure from before treatment that is readily available and was collected independent of study
- Proxy pretest design should not be actively pursued, but is rather a backup if study has begun and no proper pretest was conducted



# Separate Pre-Post Sample Design

- Two non-equivalent groups, where pretest and posttest are conducted on different subgroups each
- This setup can occur, if you can not be sure to track the same participants within each group at pretest and posttest
- I.e., trying to improve customer satisfaction: one organization is getting the treatment, another one is serving as control; customers that are surveyed during pretest will most likely not be the same as during posttest as different customers will have issues
- Variation of experiment design includes random subsampling within group – this doesn't change issues of design however

Treatment	Group 1a	Test	-	-
	Group 1b	-	X	Test
Comparison	Group 2a	Test	-	-
	Group 2b	-	-	Test
	Participants	Pretest	Treatment	Posttest

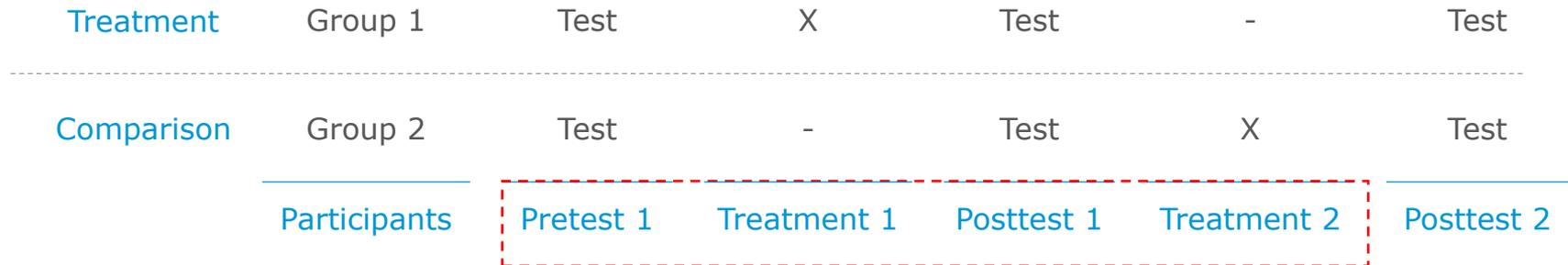
# Double Pretest Design

- Similar to classical Nonequivalent Groups Design, however two pretests are conducted
- This eliminates possible selection maturation biases, as differences in maturation could be observed between the two pretests
- Often referred to as "dry run" quasi-experimental design, as it simulates the null case
- Design is very strong in internal validity



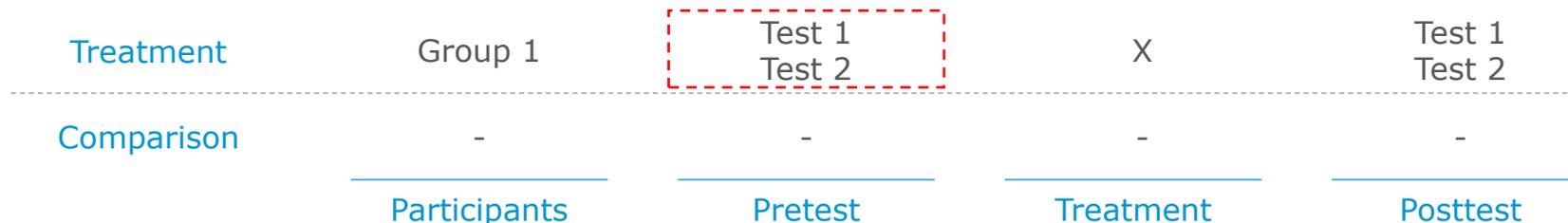
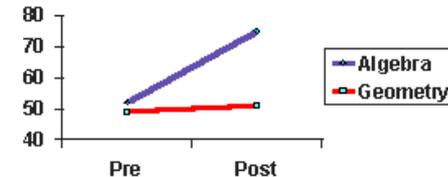
# Switching Replications Design

- Design involves two separate non-random groups, where both are treated consecutively with posttests in between
- Strong in internal validity, and strong in external validity due to two separate implementations of treatment
- Ethically strong, as all participants receive treatment at some point



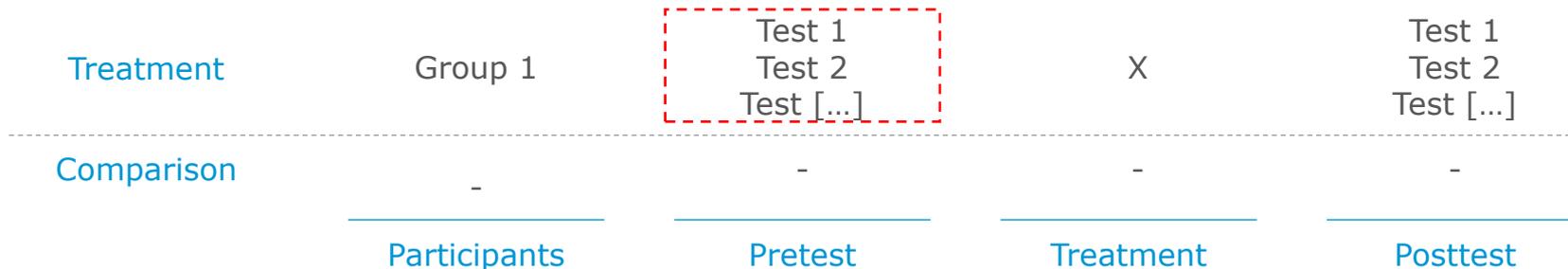
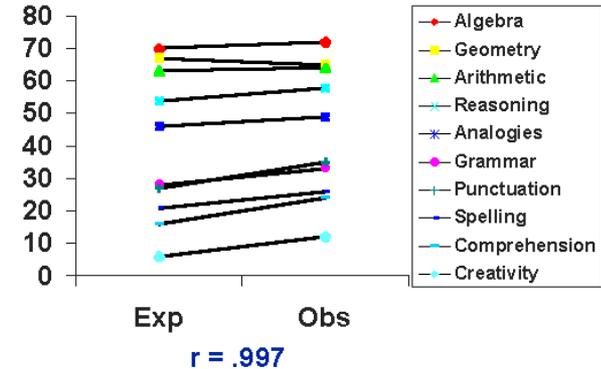
# Nonequivalent Dependent Variables (NEDV) Design

- Only treatment group (no comparison group) which is evaluated based on two different variables in pretest and posttest
- Treatment is supposed to affect one variable, but not the other variables
- Other variables serve as "control", to capture any selection threats (i.e., history, maturation, etc.)
- Key is that control variables are similar enough, so that selection threats realize identical to target variable, but not so similar that treatment affects them
- E.g., program to improve algebra score, which is controlled with geometry scores
- Overall design is rather weak in internal validity



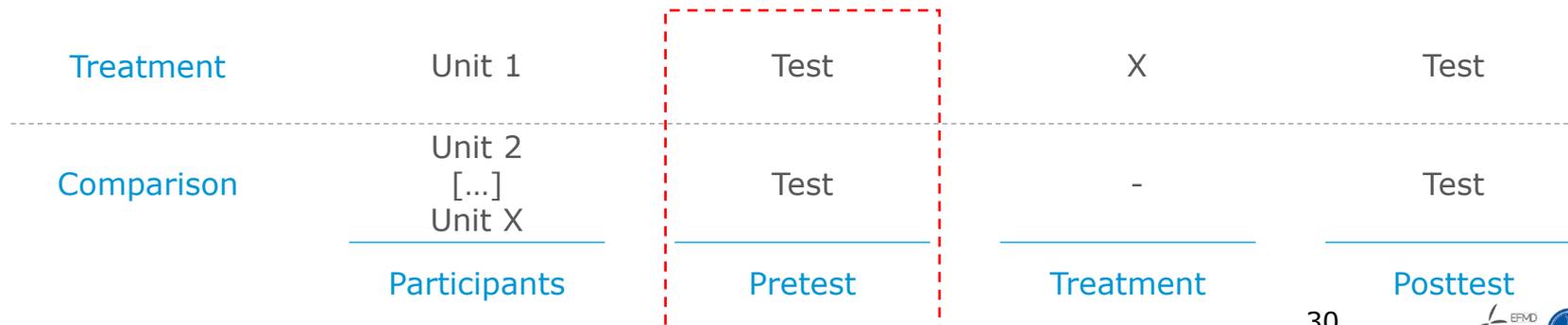
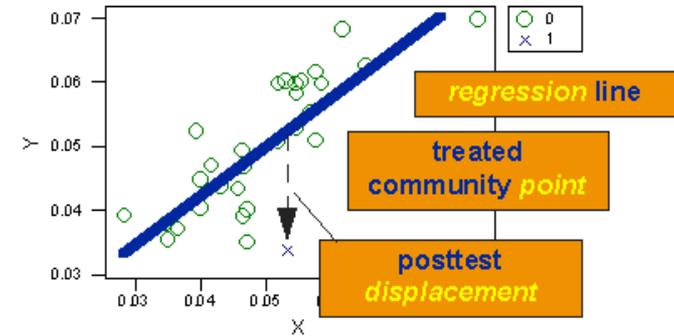
# Pattern Matching NEDV Design

- Similar to previous design, however many more test metrics are being used
- Ex ante the researcher specifies what the expected impact of the treatment is on each variable
- Matching patterns between expectation of observation is strong evidence for effect, as only other event with exactly same impact as expected could cause results
- The more variables the better internal validity, but also harder to find expected patterns in data



# Regression Point Displacement (RPD) Design

- Useful when implementation of treatment is expensive and thus only a single unit can be treated
- Design is enhanced by leveraging a larger number of comparison units
- I.e., one community receives AIDS education program; before and after HIV rates for all communities in the state are captured -> posttest displacement of treated community is observed
- Useful, when many control case units are available and routine measurement are conducted



# Today's agenda



Nonequivalent Groups Design



Regression-Discontinuity Design



Other Quasi-Experimental Designs

# Summary

## Nonequivalent Groups Design

- Uses **pre-existing groups** for control and comparison
- Relatively **easy to implement** and allows for some discretion over group selection
- Multiple **threats to internal validity**

## Regression-Discontinuity Design

- **Group assignment based on cutoff value** of pre-program metric
- Theoretically **strong internal validity**, however lower statistical power than random experiments
- Practically a few **challenges to ensure internal validity**
- **Ethically strong** as it allows treatment of those who need it most

## Other Quasi-Experimental Designs

- **Switching Replications Design**: Strong internal validity, ethically strong, however double effort for treatment
- **Pattern Matching NEDV Design**: Theoretically strong internal validity, but difficult to practically achieve
- **Regression Point Displacement Design**: Relatively strong internal validity, useful when treatment is expensive (i.e. community research)
- **Double Pretest Design**: Strong internal validity
- Proxy Pretest Design, Separate Pre-Post Sample Design,, Nonequivalent Dependent Variables Design: Not desirable but sometimes useful as least bad option

# Now some quiz!

<https://create.kahoot.it/details/511c171c-520d-4787-a9bd-8e3636de4b5d>



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