## research design

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## <u>outline</u>

intuition

research design basics

DID and program evaluation (Wheelan, 2013, ch13)

level of analysis

## <u>outline</u>

#### intuition

research design basic

DID and program evaluation (Wheelan, 2013, ch13)

level of analysis

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#### **statistics**

- tell? what the audience needs to know?
- "the science of collecting and analyzing data for the purpose of drawing conclusions and making decisions."

• "statistics is the science of learning from data"

• good data are the key! GIGO (Wheelan, 2013, ch7): eg "shy trump", drug activity, prostitution, victimless crime

• it's just storytelling! what data are telling? what i want to

- what to study?:what you're interested in (and usually knowledgable about)
- what is doable (there are realtively easily accessible data)what will further your career (think beyond graduation!)
- [sth local/work related, applied, policy relevent]

## eg: use data to disprove your convictions!

 i knew it by heart that cities are places of largest inequalities

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https://viewing.nyc/
the-new-yorkers-interactive-maps-show-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income-income
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- #google\_vignette
- but so are unequal rural areas!

  https://www.google.com/search?q=pew+inequalit+by+county&ie=utf-8&oe=utf-8
- correlation of pop siz and gini just .1

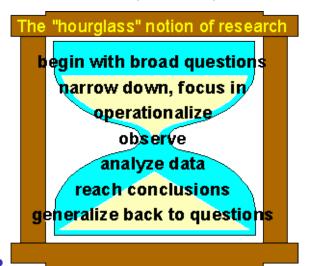
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## setup: critique res, or better yet do it yourself

- design the problem: start with a question/res idea eg?
- then formulate hypothesis(es): brief testable statement(s)
   expressed with measurable vars eg?
- get the data: download or collect/IRB (takes time/discouraged)
- summarize/analyze the data (statistics)
- interpret, communicate
- many just summarize/analyze, but need to communicate/interpret!—what does it mean?
- o interpret in the most simple way possible
- o most people don't understand statistics

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## the hourglass (Trochim)



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#### narrow down, focus in

- tendency to overcomplicate/ grand research questions
- start simple/can complicate later if resources/time
- much easier/faster to contribute locally than scholarly
- be specific about what exactly/what aspect
   YOU are looking at...
- too broad ideas cannot be tested
- o may break it down into several specific hypotheses
- anyone having any hypotheses? give me few examples?

(note how it differes from research question!)

intuition 8/56

## operationalize: have a hypothesis

- hypothesis: brief and clear statement that can be tested
- measured with variables and specified "+" or "-" effect
- express your idea in observable/measurable terms
- translate words/idea into a mathematical relationship
- eg increase in X is associated with decrease in Y
- o where X and Y are specific variables
- o say, income increases happiness
- and then use research methods, interpret results

and answer initial questions

intuition 9/56

## the trick/shortcut

- easiest way to do res is to just replicate exisiting one!
- o and add a little twist from yourself
- find a paper you really like and replicate it with a little twist from yourself :)

- sure, do follow trochim's hourglass
- but can also just do it, dive in, and and poke around

 also even if you only do qual; it does help to sprinkle it with quant!

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#### wrap-up

- end every class discussing what we covered and quick look at next week
- end with a review Q&A,
- give some examples (essp in pub pol and pub adm) for concepts covered
- students will discuss concepts from the class
- quick look at next class

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## **outline**

intuition

research design basics

DID and program evaluation (Wheelan, 2013, ch13)

level of analysis

research design basics 12/56

## qualitative vs quantitative

- much of the following applies whether you do qualitative or quantitative research
- research design is a class itself
- o we will cover only basics; for more:
- http://www.socialresearchmethods.net/kb/design.php

research design basics 13/56

## external validity (Wheelan, 2013, ch10)

- external validity is about generalizability
- o can i say something about RU in general by analyzing you?
- o how about just RU-Camden ?
- note: random sampling is different from randomization or random assignment (experiment)
- O https://sk.sagepub.com/reference/researchdesign/n146.xml
- let's have a thorough discussion like 15min, give examples, people confuse it and think it's sth more than just plain generalizability and representativness from sample to population, and often just sample=population

research design basics 14/56

## internal validity

- internal validity is about causality
- you have internal validity if you can claim that X causes Y
- o eg some drug X causes some disease Y to disappear
- O https://sk.sagepub.com/reference/researchdesign/n43.xml
- O https://sk.sagepub.com/reference/researchdesign/n192.xml

research design basics 15/56

#### causality

- much of research design is about causality
- $\circ$  want to show  $X \to Y$
- correlation is necessary for causality
- o but not sufficient (eg http://www.tylervigen.com/)
- careful! humans have illusion of causality: tend to see causality where there is none!
- https://www.ncbi.nlm.nih.gov/pmc/articles/PMC4488611
- http://onlinelibrary.wiley.com/doi/10.1348/000712610X532210/abstract

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# INUS condition (Mackie, 1980)

- very useful way of thinking about causality:
  - Insufficient but Non-redundant part of Unnecessary but Sufficient Condition
- most causes are INUS conditions
- eg a cigarette as a cause of forrest fire
- o it's Insufficient, because by itself it is not enough, eg you

also need oxygen, dry leaves, etc

- o it is contributing to fire, hence Non-redundant
- and **along with other stuff** (oxygen, dry leaves etc) it constitutes Unnecessary but Sufficient Condition
- o it's not necessary for fire, it can be lightening, etc
- o but it's sufficient it's enough to start the fire

#### **INUS** condition

- IN is your X
- US is set of X's (your X+other X's)
- the bottom line is that there are always:
- multiple alternative causes
- o and multiple steps in causal process
- or you could say there is a train of causality:
- o multiple things have to happen for outcome to occur
- say airplane fall: multiple things had to happen:
- o pilot, traffic control, weather, etc
- same with everything: career success, marriage, etc

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#### **INUS** in social science:

- nothing necessary—can have outcome with some other cause
- nonothing sufficient on its own—always need multiple things (often obviously present like oxygen in fire example)

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### basic concepts

- Y: a dependent variable, outcome
- X: an independent variable, predictor
- ∘ (T: (treatment), like X)
- Z: some other variable
- want to show  $X \to Y$ ; X affects (causes) Y
- $\circ$  and not the other way round  $(Y \to X)$
- $\circ$  and not  $Z \to Y$ ; eg X(CO<sub>2</sub>),Y(temp), Z(sun temp)
- o it is difficult to argue!
- after all, there are unknown unknowns(Z's that we are unaware of)

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## the gold standard [ask IRB appr!]

- the experimental design give few examples
- only with experiment can confidently argue causality
- and it is because randomization takes care of the known and unknown predictors of the outcome (draw a picture of 2 groups of people)
- o in other words, it establishes a counterfactual (next slides)
- but wait !
- most of the time we cannot have an experimental design because it is unethical and politically impossible eg we cannot randomly assign kids to bad school or to smoking http://www.socialresearchmethods.net/kb/desexper.php

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## causality without experiment?

- maybe, but you need to do lots of work...
- essentially you want to exclude alternative explanations
- so you act like a devil's advocate...
- and try to abolish your story / find an alternative explanation
- if you cannot find any, then your story is right ...

until disproved

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# The Problem put another way: Counterfactual esentially need to compare:

o to what would have been (Y), had the treatment not

- $\circ$  what happened to the outcome (Y) due to the treatment (T or X)
- happenedeg we got a new teacher and now kids perform better on
- to know whether the teacher caused better performance we would need to know what would have happened to SAT scores without this teacher (scores might have gone up due to Z),

o and compare it to what actually happened

SAT

## The Problem put another way: Counterfactual

- the problem is that we do not observe counterfactual (we can try to infer it though)
- counterfactual is the effect of all knowns/unknowns
- (incl. unknown unknowns)
- how do we deal with lack of counterfactual
- do an experiment!
- (or if you cannot, try to estimate or infer it somehow)

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## threats to internal validity

- can still argue causality, but think about threats!
- time: history, maturation, regression to the mean
- o things develop over time in a certain way
- selection bias, self selection
- o does smoking causes cancer?
- o maybe less healthy people select to smoke?
- something else (Z) happened that caused Y
- reverse causality
- http://knowledge.sagepub.com/view/researchdesign/n192.xml#n192

research design basics 25/56

## reverse causality OR chicken-egg dilemma

- try to find some other X that measures the same or similar concept and that cannot be caused by Y
- eg instead of education → wage; do father's education → wage (your wage can reverse cause your education, but not your father's education)
- find some exogenous (external) shock: policing↔crime
- but terror attack/alert →policing→crime; we know that policing→crime; not the other way round
- o Wheelan (2013, p227) is giving the same example!
- or dating happiness—which comes first? happy folks more likely to be dated!

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## natural experiment

by state and over time

- again most of the time you cannot have an experiment
- but there are natural experiments or exogenous shocks
- exogenous meaning that they are caused externally (like an experimenter's randomization) and somewhat randomly (at least with relation to a problem at hand
- eg earthquake (any weather, eg storm); terrorist attack;
   policy change (less random)
- any other examples of natural experiments?
  also see Wheelan (2013, p231-)
- a pretty cool one is with schooling→lexp
- o natural experiment is different min school requirement

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## examples of designing research

- say a major employer comes in,
- o say Subaru in its block group
- o or Salvation Army in its block group (Crammer Hill)
- look at housing prices (can proxy economic development)
- https://www.zillow.com/research/data/
- or gentrification, eg race by bl gr in the area https://www.policymap.com/maps
- can get many variables at bl gr level!

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>>>probably stop here and pick up next wk

or before designs shown in graphs

# ex post facto: very common; \*no\* design observational, correlational; you'll most likey do or read

- observational, correlational; you'll most likey do or read this
- we start investigation "after the fact"
- no time involved, don't know whether X precedes Y
  both, X and Y are observed at the same time examples?
- (but X must precede Y in order to be causal)
- practically impossible to argue causality here
  but cheap and big N, and good external validity
- useful, many "causes" were discovered using observational studies
- eg smoking→cancer was found out using ex post facto
- http://knowledge.sagepub.com/view/researchdesign/n145.xml
   http://knowledge.sagepub.com/view/researchdesign/n271.xml#n271

research design basics

# before-after (pre-post): blackboard: schematic

- measured Y, then do X, and then measured Y again
- eg measured readership at the library , buy some cool stats books ; measured readership again
- eg measured crime rate , put more police on the streets ;
   measured crime again
- eg measured soup consumption , changed soup ; measured soup consumption again
- anyone did pre/post? eg working at school?
- o tried new programs, new approaches?
- o or simply pre-post without T, say to identify highest and lowest gain students

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(two group) comp chng: blackboard: schematic

- eg Hypothesis: police with better guns fights crime better
- measured crime rate in 2010 in Camden and Newark
- $\circ$  in 2011 give super guns to police in Camden , (but not in Newark)
- o in 2012 measured crime rate Camden and Newark
- if crime rate dropped more in Camden than in Newark, then we have evidence that the guns worked
- o https://www.stata.com/why-use-stata/i/boxplot.png

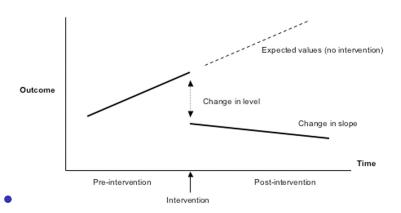
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## interrupted time series:

- eg the new anti-unemployment program in Camden decreased unemployment
- get data about unemployment in Camden 1990-2010
- say the unemployment program began in 2001
- produce a time series plot (mark a vertical line in 2001: intervention/treatment)
- if there was a change in trend after 2001, conclude the program worked

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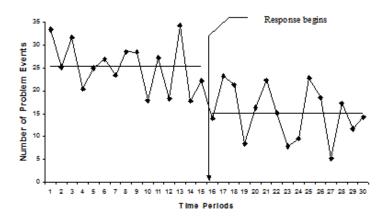
### interrupted time series:



• in general look at the trend

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## interrupted time series:



• look at the trend: may be difficult to see response

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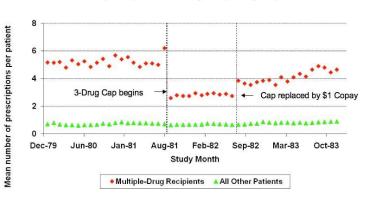


• more powerful: take away  $T \rightarrow$  effect dies

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# Interrupted time series with a control Interrupted Time Series

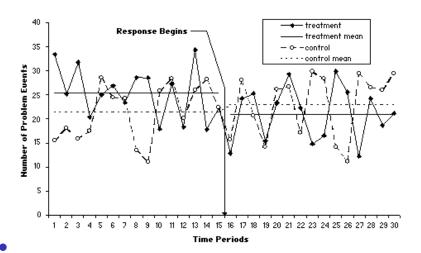
Average number of constant-size prescriptions per continuously eligible Medicaid patient per month among multiple drug recipients



Adapted from: Soumerai et al, N Engl J Med 1987

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#### interrupted time series with a control



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#### <u>outline</u>

intuition

research design basics

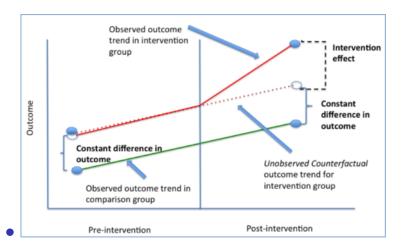
DID and program evaluation (Wheelan, 2013, ch13)

level of analysis

# difference in difference (p.235 Wheelan, 2013)

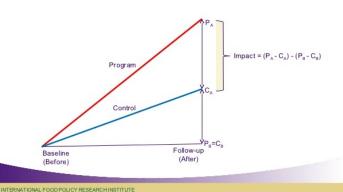
- just 'before after' with a comparison group
- did sth to one group, and not to the other group
- o over time (pre post) see if there is any difference
- like we discussed earlier in res\_des.pdf
- blackboard: fig: first from p236, and then from p237
- o and pictures similar to those from res\_des.pdf follow

#### DID





# Illustrating Difference-in-Difference Estimate of Average Program Effect



DID and program evaluation (Wheelan, 2013, ch13)

# discontinuity analysis (p.238 Wheelan, 2013)

- can use when there is some rigid cutoff for something, say:
- remedial program for F grades
- prison sentence for a crime
  then compare those who just made it (C-, or a ticket)
- o v those who didn't (F, prison)—but they were just above
- the cool thing is that the two groups are similar, especially:
- not really any difference whatsoever with respect to cause
- o so the treatment is arbitrary (random), so we have experiment!

the cutoff

of treatment!

## a general example of using res des

- new jersey state government workforce profile 2010
- http://www.nj.gov/csc/about/publications/workforce/pdf/ wf2010.pdf
- p37: minorities in state govt over time
- how increase internal validity?
- compare to PA, DE, NY etc
- factor in minority population; applications
- do experiments! many already done! again, read lit!!
- say people with black names apply for jobs
- o students with Asian names email professors
- and both, employers and professors discriminate against!

## eg: tacit knowledge is the key!

- if you know sth about state govt
- you know that it is concentrated in Trenton
- o (one student said so)
- hence, the key is population characteristics
- o around Trenton!

#### next step

- if you are interested in program evaluation:
- O quick http://www.socialresearchmethods.net/kb/evaluation.php
- o in-depth, advanced: Mohr (1995), Shadish et al. (2002)

### **outline**

intuition

research design basics

DID and program evaluation (Wheelan, 2013, ch13)

level of analysis

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#### levels of analysis

- you are familiar with term Unit of Analysis (U/A)
- there are many levels
- there are states, counties, metropolitan areas, cities, etc
- and you often get different and even opposite conclusions depending on what level you are looking at

level of analysis 48/56

#### aggregate data

- in regional development research much of the data is aggregate
- eg income, home ownership rate at county level are sums of person-level values divided by population
- with aggregate data you are losing information you don't know the variability and the distribution

level of analysis 49/56

# ecological fallacy

- it happens when you make conclusions about individual units based on group data
- eg on vacation in Hawaii you meet a person from Camden
- o and you think: "she must be a criminal"
- that Camden has the highest crime rate in the U.S. does not mean that everybody in Camden is a criminal
- now say, you meet a person that graduated from Harvard
- o and you think "she must be a genius"
- again, just because Harvard is ranked as a best university (U.S. News) does not mean that every Harvard graduate is a genius
- http://www.socialresearchmethods.net/kb/fallacy.php

#### atomistic fallacy

- an opposite of ecological fallacy
- making inferences about groups based on individual data
- eg you found that rising individual income reduces risk of coronary heart disease (eg people stress out that they are relatively poor, they are missing out...)
- but it does not mean that increasing incomes of states would decrease coronary disease rate for a state ...
- http://www.paho.org/english/dd/ais/be\_v24n3-multilevel.htm

level of analysis 51/56

## different levels, different effects

- variables at different levels may have opposite effects
- eg if i increase your salary, you'll be happier
- but if i increase salary of everybody in your county you'll be less happy
- would you like to live in a world where you make \$100k
   and the average is \$150k
- or would you like to live in a world where you make \$75k
   and everybody and the average is \$50k
- people chose the second scenario
- "a rich guy is a one who makes \$100 more than his wife's sister's husband"

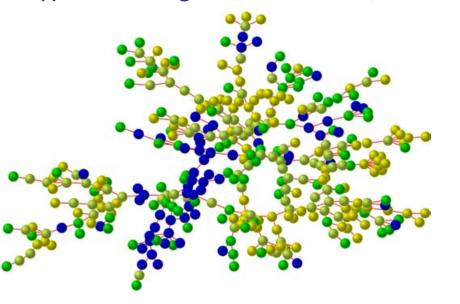
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#### contextual effects

- a closely related concept is of contextual effects
- whatever you study it takes place somewhere and place matters
- $\bullet$  so it is not only characteristics of the U/A that predict your outcome
- $\bullet$  but also the context (characteristics of larger units in which U/A is nested)
- student is nested within a classroom, a classroom within school, a school within a district, etc etc
- a firm is nested within a city/metropolitan area/town, which is nested within a state, which is nested within a country

level of analysis 53/56

### happiness is contagious (Fowler and Christakis, 2008)



level of analysis 54/56

#### your research project

- you should address some of the above issues in your research project
- again, a useful thing to do is be devil's advocate
- ask yourself how/why what you are saying is not true
- think about alternative explanations
- what are the limitations of your study

level of analysis 55/56

#### wrap-up

- end every class discussing what we covered and quick look at next week
- end with a review Q&A,
- give some examples (essp in pub pol and pub adm) for concepts covered
- students will discuss concepts from the class
- •
- quick look at next class

level of analysis 56/56

- Network: Longitudinal Analysis Over 20 Years in the Framingham Heart Study," British Medical Journal, Vol. 3, January 09.

  MACKIE, J. (1980): The cement of the universe, Clarendon Press Oxford.
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