

# Causality and Experiments

<http://philosophy.ucsd.edu/faculty/wuthrich/>

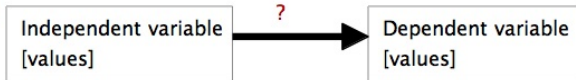
## **12 Scientific Reasoning**

Acknowledgements: Bill Bechtel

# The basic idea of an experiment

- If the independent variable is the cause of the dependent variable, then a manipulation of the independent variable should produce a change in the value of the dependent variable.
- And if it were not the cause, then we would not expect such a result from manipulation.

Manipulation



# Experiments on regular deterministic systems



- When there is no variance in the population being studied, statistical analysis is not necessary.
- The main danger is affirming the consequent: The key is to test a causal hypothesis in which you would not expect the effect to occur unless you were right about the cause.

**Manipulation**  
(vary the object dropped)

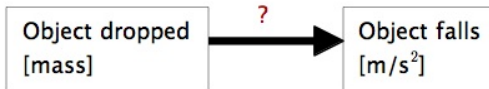


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# Variability in non-deterministic systems

- Different systems of the same kind will vary in their responses depending on
    - their particular composition (genetics, etc.)
    - their particular history, etc.
  - The same system may respond differently on different occasions.
    - E.g. Your reaction time will differ depending on
      - how much sleep you have had
      - what you have had to drink, etc.
- ⇒ Challenge: how to detect causal relations in the face of background variability

# Experiments on non-deterministic systems

Since complex systems (**biological**, **cognitive**, **social**) are not regular in their behaviour

- researchers cannot simply do an experiment on one instance and draw a conclusion about the whole population;
- must rather work with samples and draw conclusions based on statistical analysis.
  - Are the differences in the values of the dependent variable greater than expected by chance?

# Confounding variables

- The reason different individuals behave differently is that among **extraneous variables**, some might be related to the effect of interest.
- Such variables are called **confounds** and there are two kinds that are particularly important:
  - ① subject variable **confounds**: differences between subjects in the study
  - ② procedural variable confounds: differences in the way different groups are treated
- If these variables are correlated with the independent variable and are also causes of the dependent variable, the experiment is **confounded**.

# Strategies for controlling confounding variables

- 1 randomization
  - most commonly used to control confounding subject variables
- 2 locking
  - most commonly used to control confounding procedural variables
- 3 matching subjects on confounding variables
- 4 making confounding variables into studied variables



# Procedural variable confounds

When you conduct a manipulation, generally more than one thing will be changed:

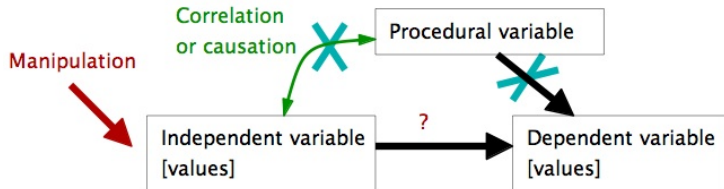
- 1 These variables will then be correlated with the independent variable—**extraneous**
- 2 If one of the other variables is causally related to the effect of interest, it rather than the variable you are considering may be the cause—**confound**

# Confounding procedural variables



- The president of the AGL corporation wanted to get her workers to be more productive.
- She found that when each employee was presented with a jar of jellybeans, productivity increased.
- Was it the jellybeans that caused the increased productivity? Or was it:
  - novelty of the situation
  - attention from the president
  - desire to reciprocate

# Controlling confounding procedural variables

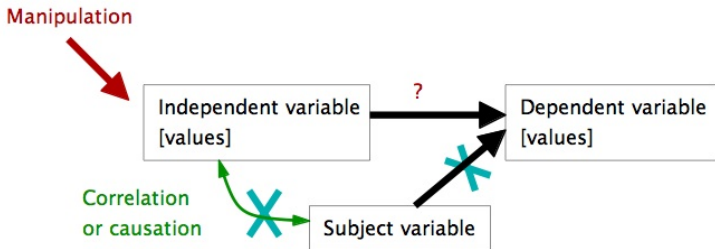


- Strategy: break the correlation—thereby breaking the effect of the confounding variable
- How to eliminate the effects of the confounding procedural variables in the jellybean case?

# Subject variable confounds

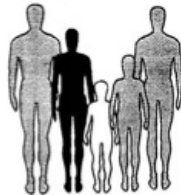
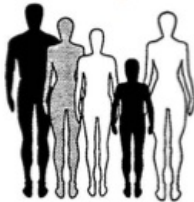
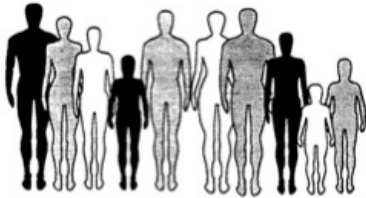
- Subjects in an experiment may be differently affected by different values of other variables.
  - people of different ages sleep different amounts
  - women might be affected differently than men
- These variables are **extraneous**.
- If there is a correlation between these variables and the independent variable, they, rather than the variable you are focusing on, may be what produce the change in the dependent variable.
- Such variables are **confounds**.

# Controlling confounding subject variables

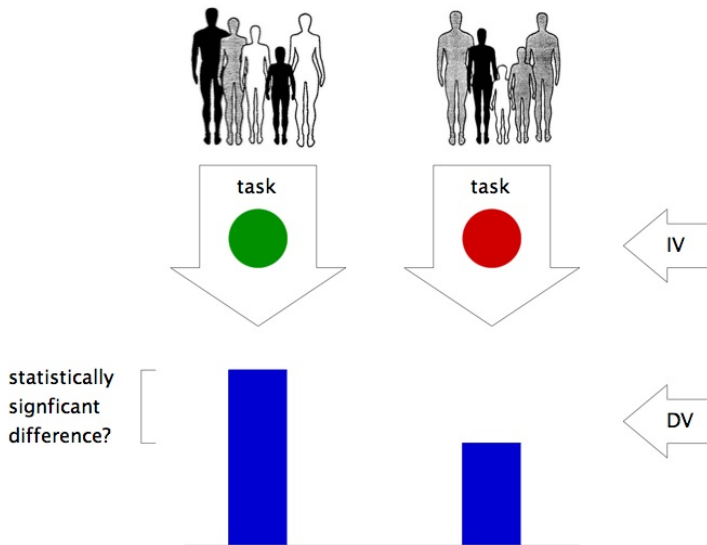


- Strategy: break the correlation—thereby breaking the effect of the confounding variable
- random assignment of subjects is a strategy for breaking the correlation

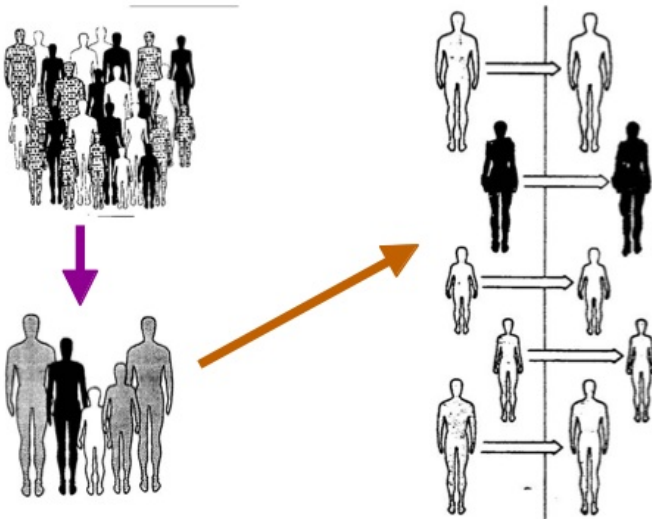
# Controlling subject confounds: Between subjects randomization



# Manipulate independent variable

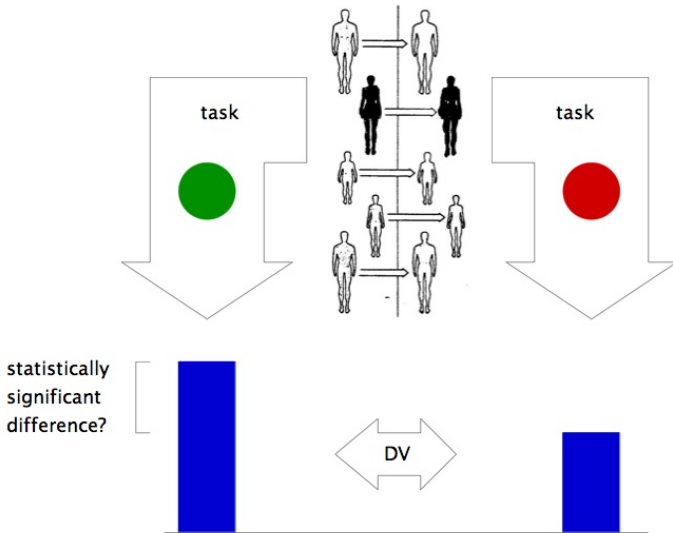


# Controlling subject confounds: Within subjects designs





# Subjects serving as own control



# Between-subject design

## Good news:

- Subjects are not 'contaminated' in one condition as a result of having participated in the other.

## Bad news:

- requires a larger number of participants
- runs the risk of non-equivalence of subject groups

# Within-subject design

## Good news:

- requires a smaller number of participants
- rules out any differences between subjects

## Bad news:

- potential 'contamination' of participants' behaviour from previous trial: **carryover effect**
- subjects might learn from one condition and that could alter their behaviour in the second condition
  - practice effect
  - fatigue effect

# Counterbalancing

## (1) within-subject counterbalancing

- reversing order: ABBA

## (2) across-subject counterbalancing

- complete: every possible sequence—requires  $n!$  runs
- partial
  - random
  - Latin square:
    - each condition appears once and only once in a given ordinal position
    - no two conditions are juxtaposed in the same order more than once

Order 1	A	B	D	C
Order 2	B	C	A	D
Order 3	C	D	B	A
Order 4	D	A	C	B

# Pretest-posttest design

- There is always a danger in an experiment that the members of the two (or more) groups being studied already differ on the dependent variable.
- Best control is to focus on change, not raw value of the dependent variable.
  - **Pretest**: measure the dependent variable before the intervention
  - **Posttest**: measure the dependent variable after the intervention
  - **Change** = Posttest – Pretest

Manipulation



Independent variable  
[values]

?



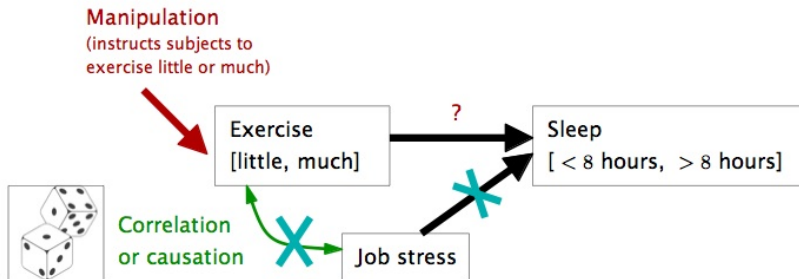
Dependent variable  
[change]

# Limitations of pretest-posttest design

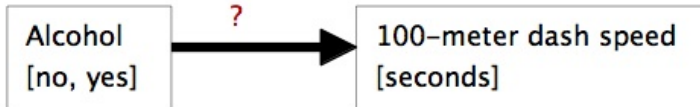
- Just measuring change in one group using a pretest and a posttest allows for confounds:
    - Time has elapsed and subjects have gotten older (maturation).
    - Events occurring between the pretest and posttest could affect the dependent variable (history).
    - Experience with previous test may change performance.
    - Pretest and posttest may vary in difficulty.
- ⇒ Use of pretest-posttest does not obviate the need for a control group.

# Example: exercise and sleep

Is there a causal relation between exercise and sleep?



# Example: alcohol and running speed





# Between subjects or within subjects

Between-subjects design:

- Different subjects would be used for the no-alcohol and alcohol condition, and each would be tested only once.

Within-subjects design:

- Each subject would be tested both under the no-alcohol and alcohol condition.

# Between subjects



Jim  
Angela  
Megan  
Tony



Roger  
Shane  
Sara  
Jessica

# Within subjects



Jim  
Angela  
Megan  
Tony



Jim  
Angela  
Megan  
Tony

# Counterbalance

Alcohol condition  $\Rightarrow$  rest  $\Rightarrow$  No-alcohol condition

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Jim  
Angela

Jim  
Angela

No-alcohol condition  $\Rightarrow$  rest  $\Rightarrow$  Alcohol condition

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Megan  
Tony

Megan  
Tony

# Internal validity

## Definition (Internal validity)

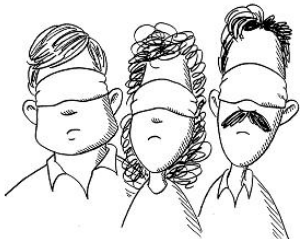
*An experiment is **internally valid** if it was in fact the manipulation of the independent variable that produced the change in the dependent variable*

- Are the effects on the dependent variable due solely to the manipulation of the independent variable?
- Was there a **confounding subject variable** that did not get controlled?
- Was there a **confounding procedural variable** that did not get controlled?

# A threat to internal validity

- People may change their behavior when they are being observed:
  - People want to be liked (or not!).
  - People want to be helpful (or not!).
  - People want you to think they're a good person.
  - People want to be thought of as intelligent and normal (not crazy, stupid or obsessed).

⇒ Problem if subjects figure out the point of an experiment.



- Solution: keep subjects blind as to the point of the experiment or what is being studied (**single-blind experiment**)

# The legend of the Western Electric Plant at Hawthorne



- test of improved working conditions on worker productivity
- legend has it that any change that was made increased worker productivity:
  - increase lighting  $\Rightarrow$  work harder
  - decrease lighting  $\Rightarrow$  work harder
- no published report and the data 'has vanished'

# Hawthorne mica-splitting experiment

- Experimenters monitored the output of five experienced women who split, measured, and trimmed mica chips used for insulation at their regular department workstations.
- Experimenters moved the women to a special test room and added 10-minute rest breaks at 9:30 am and 2:30 pm.
- After a brief decline in performance, the women's output increased by an average of 0.15 percent and remained at that level.
- When the women returned to their department and lost the rest periods, their output dropped back to the original rate.
- Since they thought no other conditions had changed, the researchers attributed the increase in output to the beneficial effects of rest periods—not an example of the [Hawthorne effect](#).
- And there were confounds aplenty in this study.



# Experimenter bias

- Danger that experimenters will see what they want to see.
  - Mendel's data is too perfect—there should be more variability
  - most likely explanation is that he did not deliberately cheat (remember, he was a monk!)
  - but he reported the best cases and subjectively biased his counting of plants
- ⇒ keep the data-tabulator blind as to which group different subjects are in
- ⇒ double-blind study

# Planning an experiment

Say the colour the following words are written in:

**Blue**

**Pink**

**Brown**

**Yellow**

**White**

**Orange**

**Red**

**Green**

- Does it seem harder to name the colour when the words name a different colour?

How might we test the claim that it is the meaning of the word that makes it harder to say the colour it is written in?

- 1 operationalize the notion of being hard to read
  - slower reaction time when incongruent words
- 2 identify a sample population
  - college undergraduates in psychology courses
- 3 pick study design
  - between subject
  - within subject

# Controlling subject variable confounds

Blue

Pink

Brown

Yellow

White

Orange

Red

Green

- What subject variables might you have to worry about as confounds?
- How to control for these confounds?
  - if between subject  $\Rightarrow$  randomize
  - if within subject  $\Rightarrow$  counterbalance

# Controlling for procedural variables

Blue

Pink

Brown

Yellow

White

Orange

Red

Green

- What procedural variables should be controlled to avoid confounds?
    - context of presentation
    - illumination of the stimuli
    - length of words
    - familiarity and frequency of words
- ⇒ need to lock these variables so that they do not vary across conditions

# Find two major confounds

Test for verbal (lecture) versus visual (reading) learners

- in a different context, have students self-identify as verbal versus visual learners
- students then tested for memory of 20 different words presented orally (tape recorder) or visually (paper)
- lists were presented first in the subject's preferred format, then other
- oral and visual lists the same for all subjects
- all other variables (length of presentation, delay before memory test held constant)

# Detecting pseudo-causal relations



- Can **dowsers** reliably locate water or metal objects?
- Ray Hyman tested an experienced dowser on the PBS program *Frontiers of Science* (19 Nov 1997):
  - used random numbers to pick which of the buckets that were placed upside down in a field would have a metal object under it
  - no one going with the dowser knew which buckets had been selected

- If there are 100 buckets and 10 of them have a metal object, then getting 10% correct would be expected by chance.
- The dowser walked up and down the lines of buckets with his rod but complained that he couldn't get any strong readings.
- Occasionally he selected a bucket but qualified his selection by expressing doubt that he'd be right.
- He was right about never being right—he didn't find a single metal object despite several attempts...
- The dowser was genuinely surprised by his failure.



# A miracle device?



Detect human beings hidden in building or behind objects from almost two football fields away...

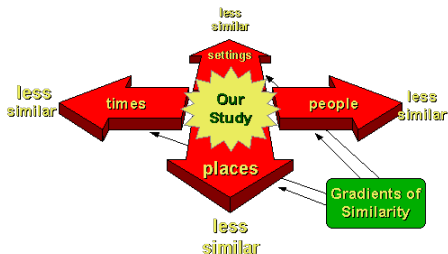
# The DKL LifeGuard

- according to the company can detect human heartbeat
- range of up to 500 meters (540 yards)
- no effective electronic or other countermeasures
- no natural and synthetic material it cannot penetrate
- no battery or any power sources required!
- repeatability of 99%
- can penetrate:
  - 10-meter wide earthen barrier
  - 10 feet of water
  - concrete walls, steel bulkheads
- can distinguish a man, woman or child from any other animal, even a gorilla or an orangutan
- requires only one day of operator training
- work as well in darkness as in daylight
- like no other technology on the market today... literally in a league of its own

# Double-blind test

- five large plastic packing crates were set up in a line a 30-foot intervals
- a DKL representative, using the DKL LifeGuard Model 2, tried to detect which of the five crates contained a human being
- On preliminary tests in which the operator knew which crate contained the person, the DKL LifeGuard found the person 10 out of 10 times.
- In the real, double-blind test, the operator found the person 6 out of 25 times (and took much longer to find the person).

# External validity



- To what extent can you **generalize** the results of your study?
- Are they specific to a particular sample?
  - college sophomores or the general population
- Do they only apply in a particular (laboratory) setting?
- Do they generalize beyond the details of the manipulation?
  - ecological validity

# Population generalization

## Question

*Will a study using one population generalize to another population?*

- Will a study of college sophomores generalize to middle-aged adults?
- Will a study of chronically depressed patients generalize to patients who are acutely depressed?
- Will a study of captive raised dolphins generalize to wild dolphins?
- Will a study on mice generalize to humans?

# Setting generalization

## Question

*Will a study conducted in one laboratory or clinical setting generalize to the setting of interest?*

- Will results obtained in a flight simulator generalize to an actual cockpit?
- Will results obtained in an outpatient setting generalize to a psychiatric hospital?
- Will results obtained in a laboratory generalize to customers in a store?

# Manipulation generalization

## Question

*Will a result obtained with one task generalize to other tasks or stimuli?*

- Will studies of perceiving visual illusions generalize to perception of ordinary objects?
- Will a survey of consumer attitudes generalize to consumer behaviour?

# Assessing external validity

Must make a plausibility judgment in assessing external validity:

- Is the target population different from the studied population in ways that are likely to matter for the causal claim?
- Is the target setting different from the studied setting in ways that are likely to matter for the causal claim?
- Is the manipulation used in the experiment different from the target process in nature in ways that are likely to matter for the causal claim?



# Example: rats and saccharine



- 1977 Canadian study which fed pregnant rats up to 20% of their body weight per day in saccharine showed an increase in bladder tumors
- Saccharine was banned in Canada and the FDA was about to ban its use in the US when Congress intervened
- Assessing external validity:
  - Are rats relevantly like humans?
  - Is eating in the laboratory like eating at home, etc.?
  - Is feeding up to 20% of body weight like eating as part of diet?