

The Logic of Research

Soc 357

Fall 2006



General Form of a Proposition

- ***Conceptual***: For population (P) in condition (C), independent variable (X) causes dependent variable (Y)
- ***Operational***: For sample (p) in condition (c), independent variable (x) has a statistical association with dependent variable (y)

Induction

- ***Induction***: reasoning from the specific to the general = Empirical generalization
- No logical proof of induction: future cases may be different from those you have seen
- ***Sampling theory*** tells us how we can use the observations we have made to make probabilistic statements about future cases

Induction Examples - 1

- This ice is cold
All ice is cold
- Teenagers get lots of speeding tickets
All teenagers speed

Induction Examples - 2

- 62 percent of voters in a random sample of 400 registered voters (polled on February 20, 2004) said that they favor John Kerry over George W. Bush for President in the 2004 Presidential election.
- This supports the generalization that between 57 percent and 67 percent of **all** registered voters favor Kerry over Bush for President (at or around the time the poll was taken)

Use of Induction in Social Science

- Field observation, exploratory research uses induction in order to ***build*** theories
- Use induction in sample studies, where we generalize about a population based on a smaller group

Deduction

- ***Deduction***: reasoning from the general to the specific, following the rules of logic
- Used to ***test*** theories. But we have to be careful about what conclusions we draw.

*Ex. 1: When it rains, the ground gets wet
 It is raining
 Therefore the ground is wet.*

*Ex. 2: When it rains, the ground gets wet
 The ground is wet
 Therefore it is raining.*

Example of Illogic of “Proof”

- Theory: Students who participate in extra-curricular activities are more likely to vote
- Data: A survey shows student athletes are more likely to vote than non-athletes
- Does this mean that your theory is true?

Illogic: Data “Proves” Theory

If theory is correct, then X is true.

X is true.

Therefore, theory is correct.

INVALID LOGIC: *X might be true for another reason*

Logic of Falsification

If theory is correct, then X is true.
 X is not true.
Therefore, theory is not correct.

VALID LOGIC

Example with logic of falsification

- Theory: Students whose parents talk politics at the dinner table are more likely to vote
- Data: A survey shows students whose parents talk politics at dinner are no more likely to vote than students whose parents talk about non-political stuff at the dinner table
- Can we conclude that the theory is false?
YES (assuming reliable & valid measures)

Falsification

- We cannot prove theories to be correct
- We CAN prove theories to be INCORRECT
- Research proceeds on a logic of falsification
 - We subject theories to tests which could falsify them
 - If a theory avoids falsification, we say it is “confirmed” (not proven)
 - If a theory repeatedly avoids falsification, we build our confidence that it is correct, but it could still be proven wrong later

The Logic of Research

- We try to falsify other theories while confirming our own

Theory A: Participating in activities → higher voting rates among youth

Theory B: Parental Socialization → higher voting rates youth

Gather data that confirms Theory A (student athletes & non-athletes data) while falsifying Theory B (parents who talk politics vs. parents who don't talk politics)

Causation

- It is impossible directly to observe causation
- Criteria for theorizing about causation:
 - **Statistical association:** two things vary together
 - **Direction of influence:** Cause precedes effect in time
 - **Elimination of rival hypotheses:** Other variables are ruled out as possible explanations for the relationship
 - **We can identify the mechanism** for the cause-effect relationship, we know how it works

Full Logic of Hypothesis Testing

Research Syllogism:

If A causes B {theory}

And if X measures/indicates A
{measurement assumption}

And if Y measures/indicates B
{measurement assumption}

Then X will be statistically associated with Y
{prediction}

Confirmation of Theory

Research Syllogism:

If A causes B {theory}

And if X measures/indicates A {measurement assumption}

And if Y measures/indicates B {measurement assumption}

Then X will be statistically associated with Y {prediction}

Data 1:

X is statistically associated with Y {prediction is correct}

We cannot prove that A is associated with B, but we can say the data *confirms* or *supports* our theory that A causes B (also confirms measurement assumptions)

Disconfirmation of Theory

Research Syllogism:

If A causes B {theory}

And if X measures/indicates A {measurement assumption}

And if Y measures/indicates B {measurement assumption}

Then X will be statistically associated with Y {prediction}

Data 2:

X is NOT statistically associated with Y {prediction is wrong}

Then at least one assumption must be wrong. Either

A does not cause B {theoretical assumption is wrong} AND/OR

X does not measure A {measurement assumption is wrong}

AND/OR

Y does not measure B {measurement assumption is wrong}

But: Falsification may show up due to sampling error or extraneous variables – more on this later

Statistical Association

- A change in one variable is associated with a change in another variable, in a way that is not likely to have occurred just by chance

Statistical Association

- There are “tests of significance” that we will not do
- Rule of thumb: association doesn’t change direction if one or two people changed responses
- Instead, we’ll focus on three general outcomes
 - Confirms the hypothesis
 - Disconfirms the hypothesis
 - Indeterminate

*** The most common mistake is to call a zero relationship “indeterminate”

Types of Statistical Association (Bivariate)

Independent	Dependent	Statistic
Qualitative	Qualitative	Difference of Conditional Percentages
Qualitative	Quantitative	Difference of Means
Quantitative	Quantitative	Correlation or regression

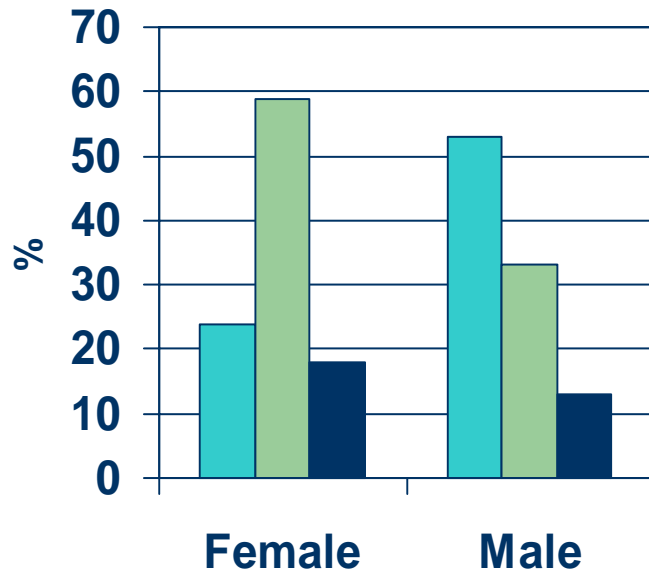
Difference of Conditional Percentages: Sex and Ice Cream Cone Eating

	Male	Female
Bite	53%	24%
Lick	33%	59%
Other	13%	18%
Total	99%*	101%*
(N)	(15)	(17)

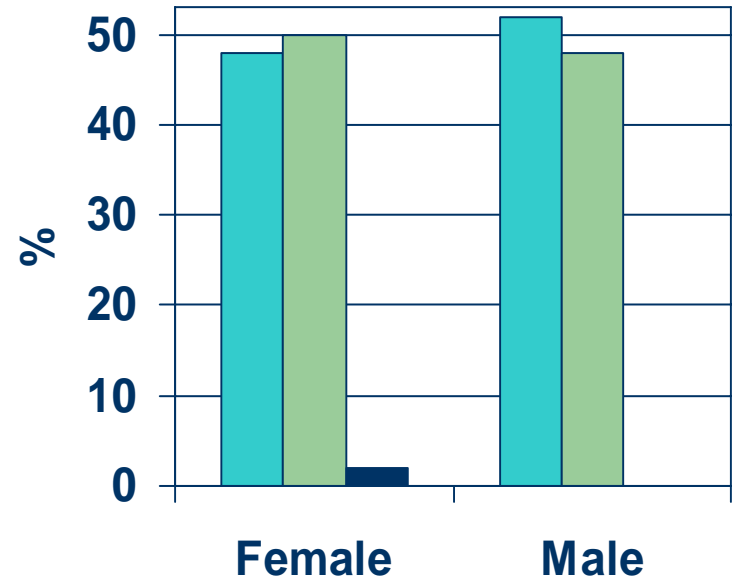
Statistical Association:

- **Males** bit 53% of the time compared to 24% of the women (a percentage difference of 29%)
- **Females** licked 59% of the time compared to 33% for males (a percentage difference of 26%)
- “**Other**” was only slightly different for men and women.

Qualitative relationships: Association & Non-Association



■ Bite ■ Lick ■ Other



■ Bite ■ Lick ■ Other

Difference of Conditional Means: Sex and Time to Complete Sales Transactions

	Men	Women
Mean Seconds for Transaction	27.1	40.5
(N)	(20)	(27)

Interpretation: Women took 13.4 seconds longer than men, on average, to complete their sales transactions.

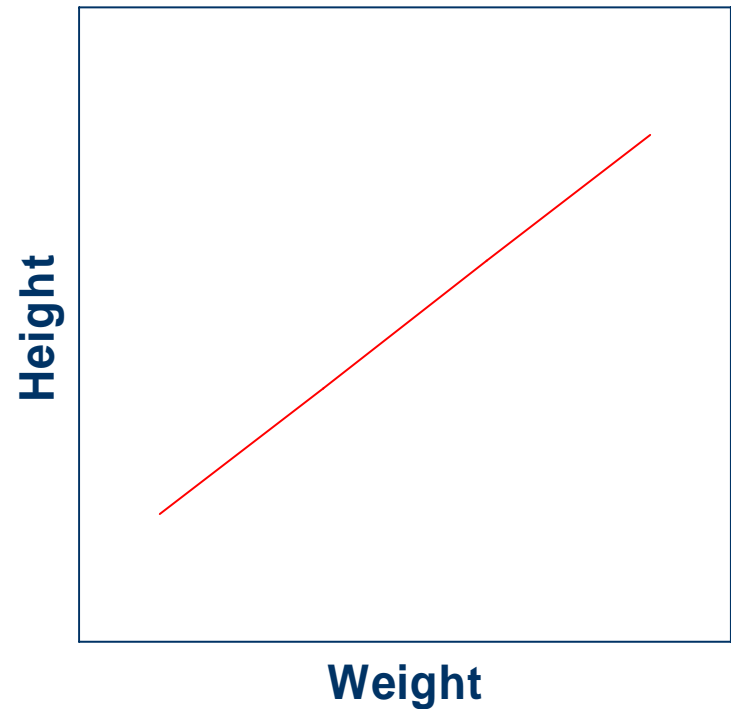
Correlations

- ***Correlations:*** calculating if a change in one variable is associated with a change in another variable
- Range between -1 (perfect negative correlation) to $+1$ (perfect positive correlation).

See: <http://noppa5.pc.helsinki.fi/koe/corr/cor7.html>

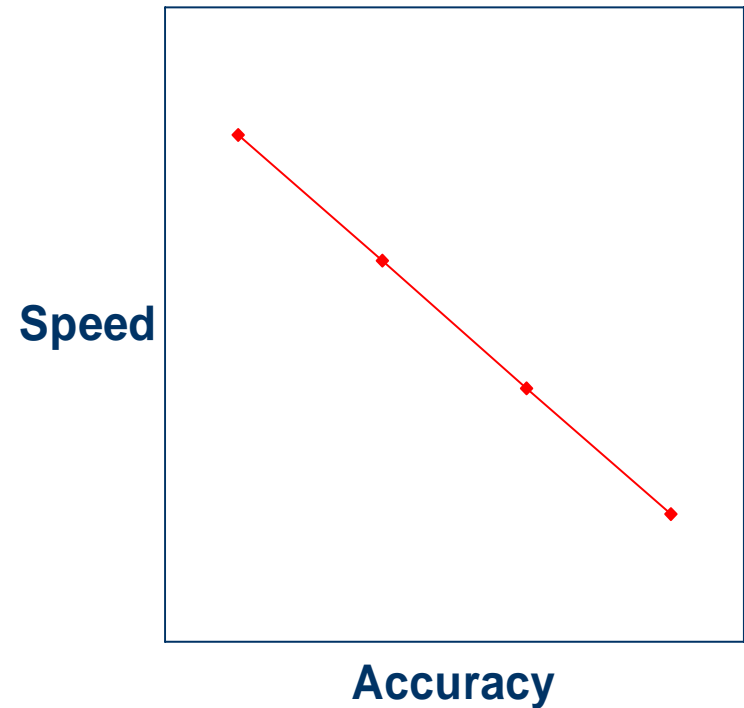
Quantitative relationships – 1

- **Positive:** when one variable is greater, the other tends to be greater
- Eg. Height is *positively* associated with weight. The taller you are, the more you are likely to weigh.



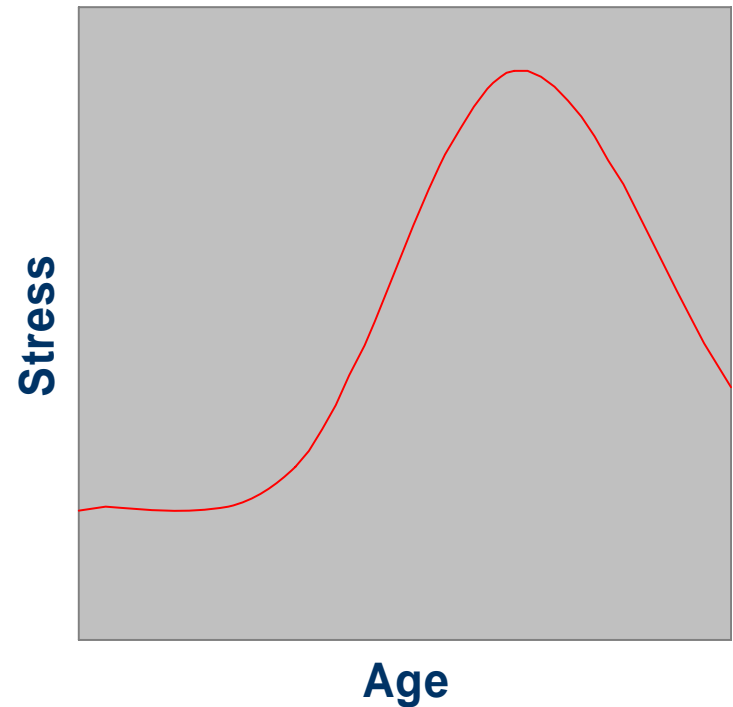
Quantitative relationships – 2

- **Negative:** When one variable is greater, the other tends to be smaller
- Eg. Speed is *negatively* associated with accuracy. The more you rush, the worse your accuracy is.



Quantitative relationships - 3

- **Curvilinear:** Any non-linear relation, but especially one that is first positive and then negative, or vice versa
- Eg. Stress is related *curvilinearly* to age. Middle aged people feel the most stress, while young & old report less stress.



Spuriousness

- Problem: A statistical relationship between X and Y may be the result of some other variable, Z.
- Solution: hold extraneous variables “constant”
 - We check that the relationship still exists for different groups of variable z
 - E.g. Smoking, gender, and lung cancer
- Problem: Can't control for everything