Geographic Data Science -Lecture IX

Causal Inference

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Today

- Correlation Vs Causation
- Causal inference
- Why/when causality matters
- Hurdles to causal inference & strategies to overcome them

"Association breeds similarity" (sometimes)

Nasir bin Olu Dara Jones (a.k.a. *Nas*)

Two fundamental ways to look at the relationship between two (or more) variables:

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Correlation

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Causation

There is a "cause-effect" link between the two and, as a result, they display co-movement.

- Both are useful, but for different purposes
- Causation implies correlation but not the other way around
- It is vital to keep this distinction in mind for meaningful and credible analysis

Sign correlation? Causal link?

- Temperature and ice-cream consumption
- Non-commercial space launches & Sociology PhDs awarded
- Crime & policing
- IMD Moran Plot in Liverpool

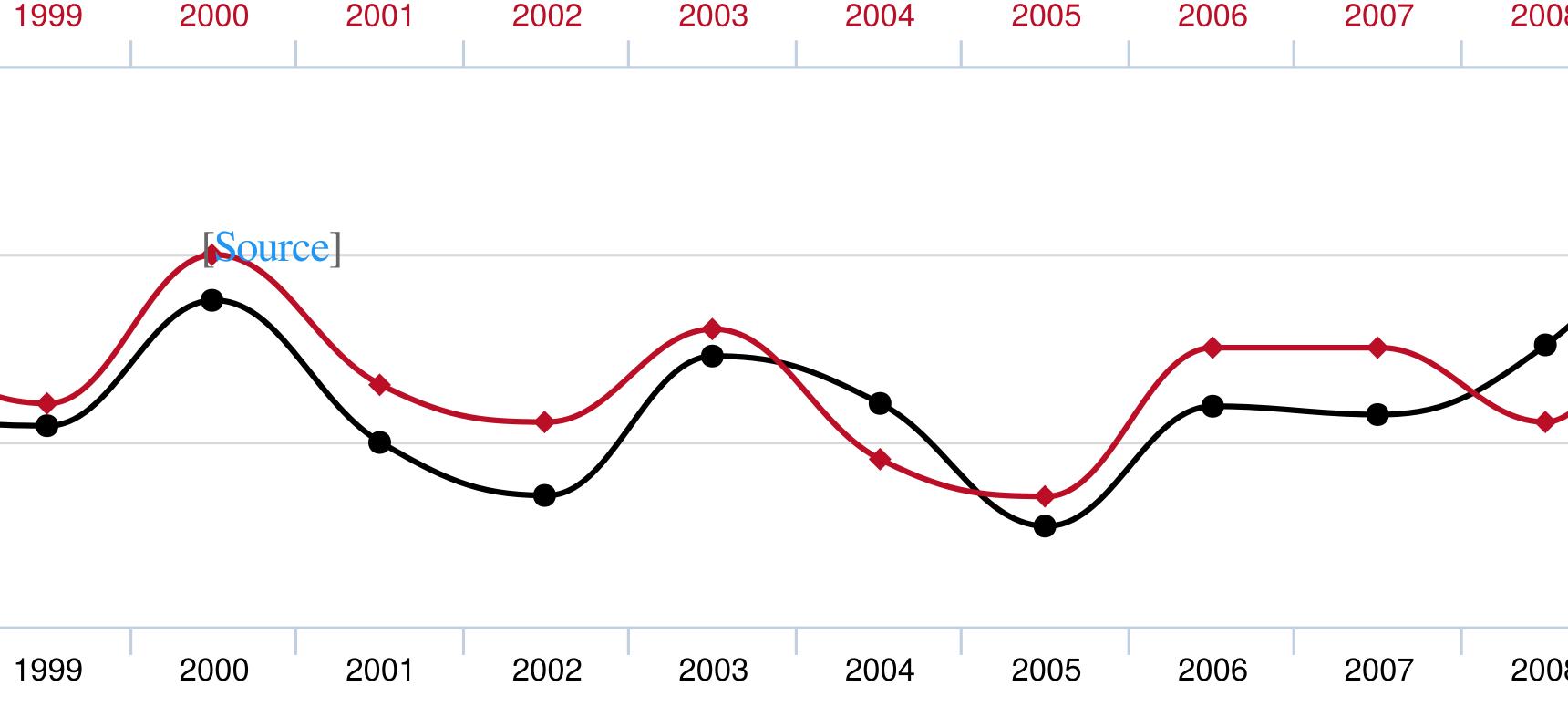
Sign correlation? Causal link?

- Temperature and ice-cream consumption \rightarrow **Positive**. **Positive**.
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Worldwide non-commercial space launches

correlates with

Sociology doctorates awarded (US)

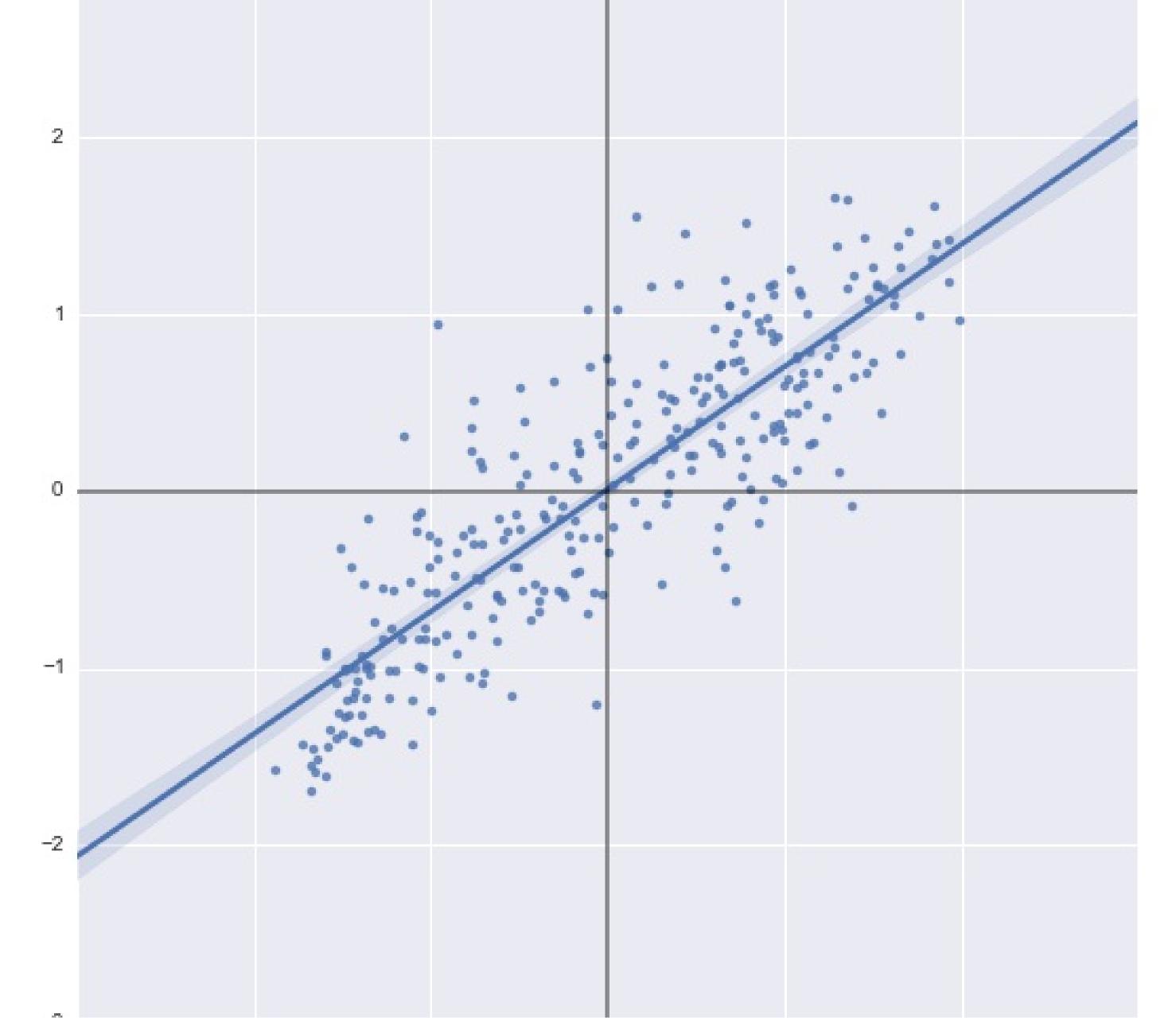


Positive or negative correlation? Causal link?

- Temperature and ice-cream consumption \rightarrow **Positive**. **Positive**.
- Non-commercial space launches & Sociology PhDs awarded →
 Positive. None.
- Crime & policing
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Positive or negative correlation? Causal link?

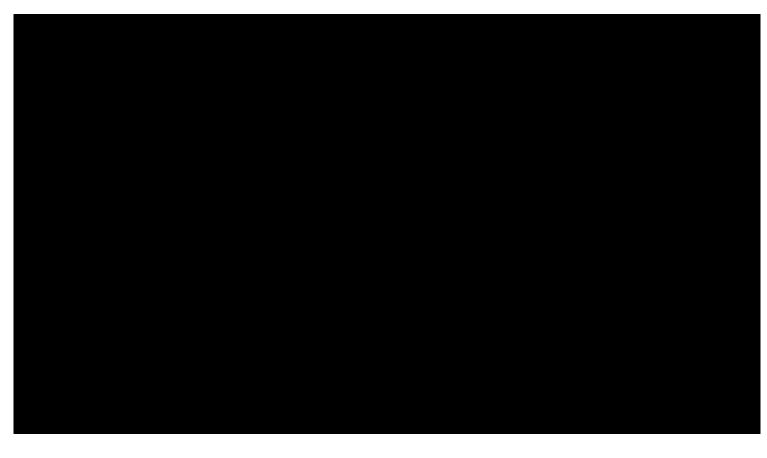
- Temperature and ice-cream consumption \rightarrow **Positive**. **Positive**.
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 Positive. None.
- Crime & policing \rightarrow Positive. Negative.
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Positive or negative correlation? Causal link?

- Temperature and ice-cream consumption \rightarrow **Positive**. **Positive**.
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 Positive. None.
- Crime & policing \rightarrow Positive. Negative.
- IMD Moran Plot in Liverpool → **Positive**. ?

Causal inference



[Source]

Why/When get causal?

Why

- Most often, we are interested in understanding the **processes** that *generate* the world, not only in observing its outcomes
- Many of these processes are only indirectly observable through outcomes
- The only way to link both is through causal channels

When

Essentially when the **core interest** is to find out if **something** *causes* **something else**

- Policy interventions
- Medical trials
- Business decisions (product/feature development...)
- Empirical (Social) Sciences
- ...

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Exploratory analysis

When you are not sure what you are after, inferring causality might be too high of a price to pay to get a sense of the main relationships

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Predictive settings

Interest not in understanding the underlying mechanisms but want to obtain best possible estimates of a variable you do not have by combining others you do have

E.g. Population density in a specific point using population density in all available nearby locations

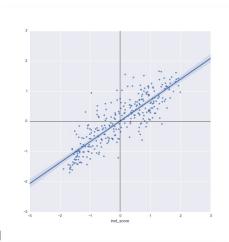
Correlation implies Causation

Causation does not imply Correlation

Why?

Correlation implies Causation

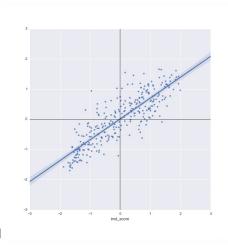
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Why?

Correlation implies Causation

Causation does not imply Correlation



Why?

- Reverse causality
- Confounding factors/endogeneity

Reverse causality

There *is* a causal link between the two variables but it either runs the oposite direction as we think, or runs in both

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E.g. Education and income

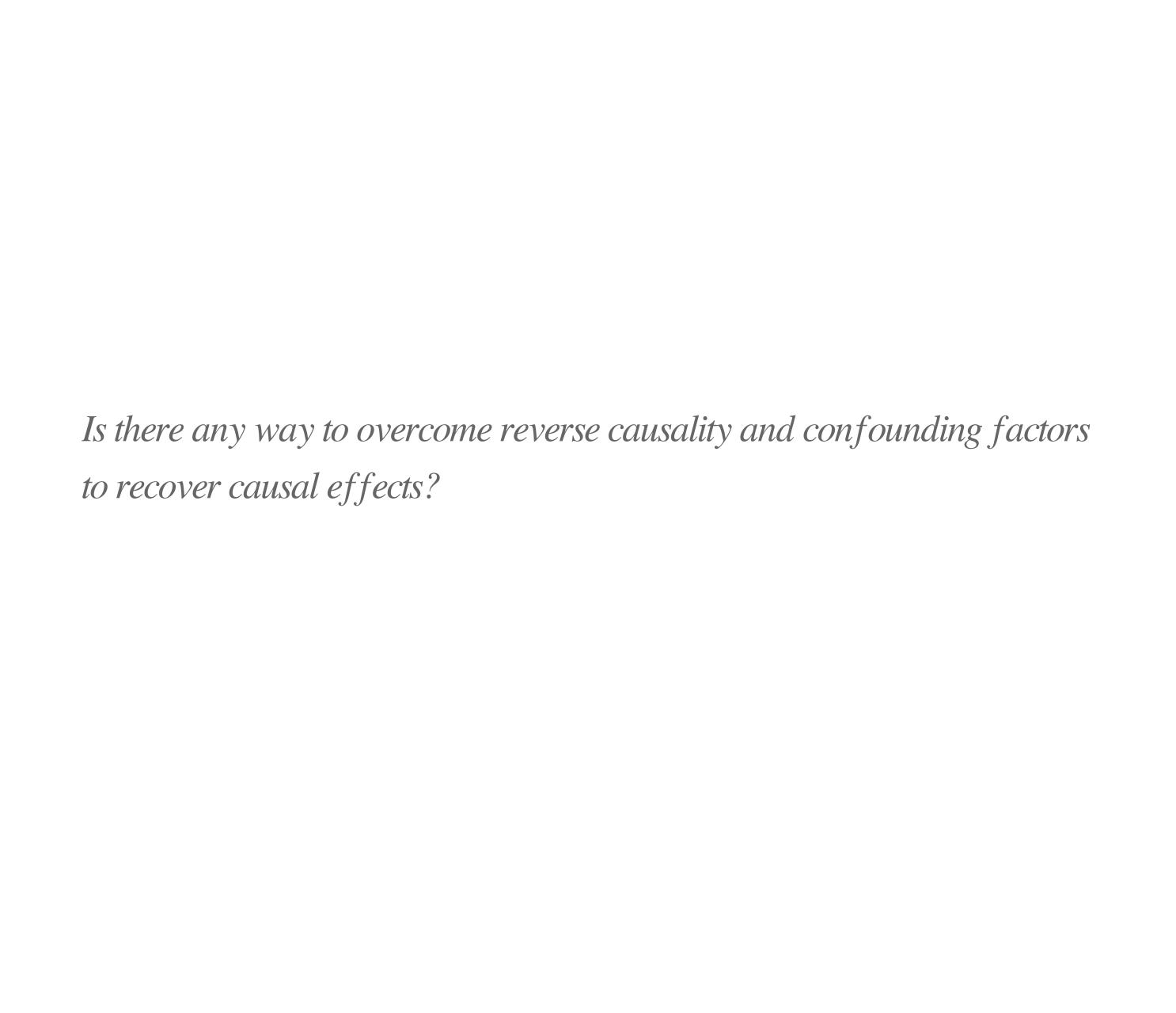
Confounding factors

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E.g. Ice cream and cold beverages consumption



Is there any way to overcome reverse causality and confounding factors to recover causal effects?

The key is to get an exogenous source of variation

Randomized Control Trials

Treated and control groups

Probability of treatment is independent of everything else

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Econometric techniques

For the interested reader: space-time regression, instrumental variables, propensity score matching, differences-in-differences, regression discontinuity...

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... correlation most often *precludes* causation and, depending on the application/analysis, it is all that is needed.

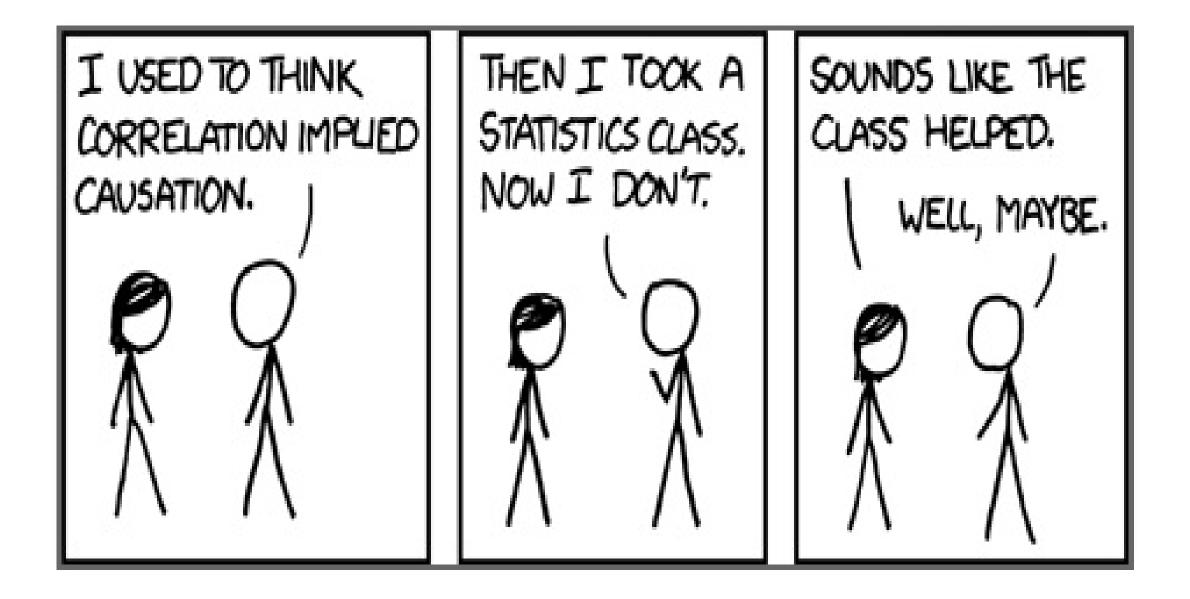
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... correlation most often *precludes* causation and, depending on the application/analysis, it is all that is needed.

It is important to always draw **conclusions based on analysis**, know what the data can and cannot tell, and stay **honest**.

Recapitulation

- Correlation does NOT imply causation
- Causality implies more than correlation, a direct **effect channel** that is **harder** to identify but might be **worthwhile**
- There are several techniques to identify causality, all usually based on obtaining exogenous sources of variation
- You don't always need causality



[Source]



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