# Geographic Data Science -Lecture IX Causal Inference Dani Arribas-Bel

# Today

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

### **Correlation Vs Causation**

### "Association breeds similarity" (sometimes) Nasir bin Olu Dara Jones (a.k.a. *Nas*)

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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.

### **Correlation Vs Causation**

- 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?

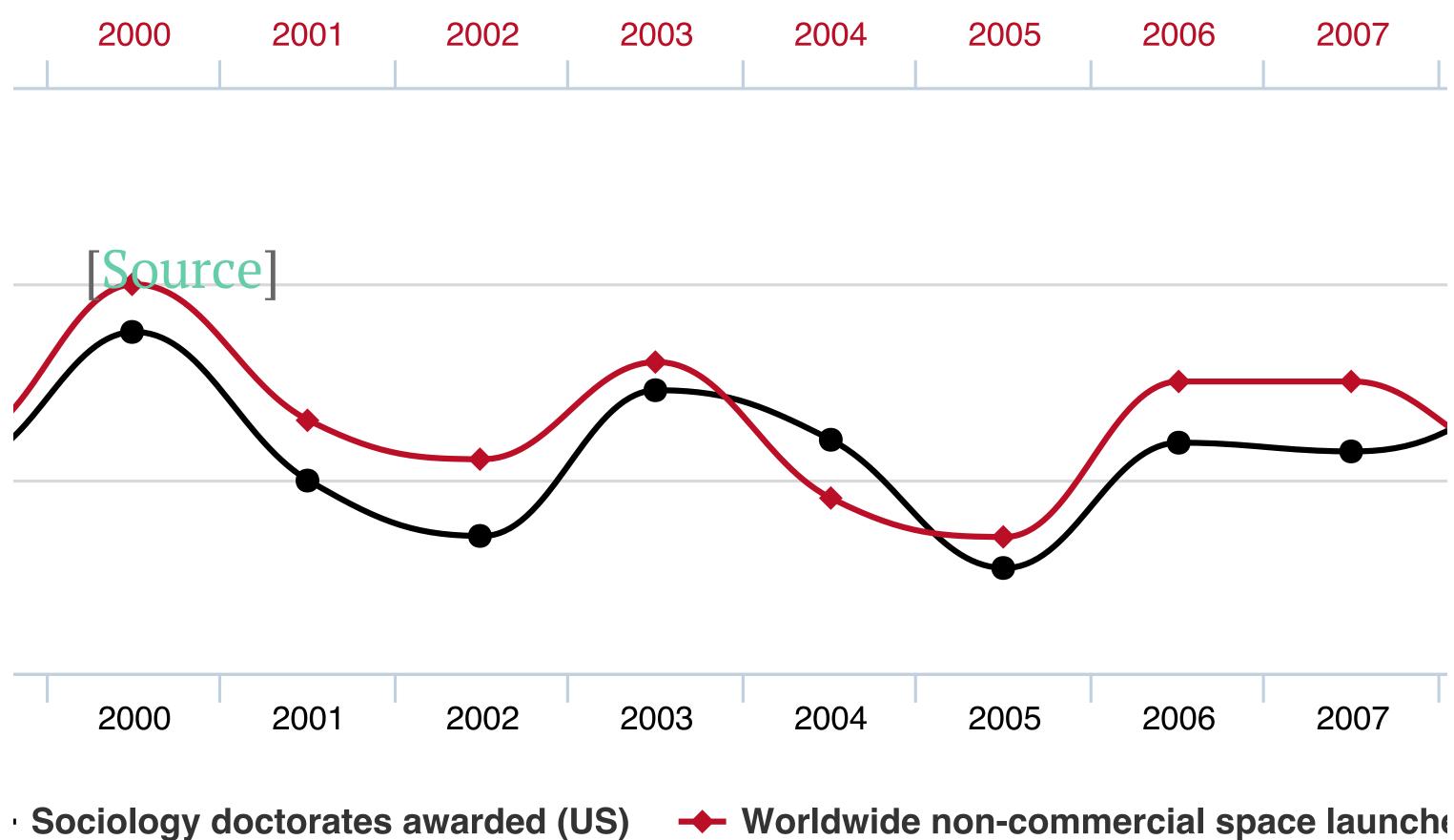
- Temperature and ice-cream consumption
- Non-commercial space launches & Sociology PhDs awarded
- Crime & policing
- IMD Moran Plot in Liverpool

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- Temperature and ice-cream consumption  $\rightarrow$ **Positive.** Positive.
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### **Norldwide non-commercial space launches** correlates with

### Sociology doctorates awarded (US)



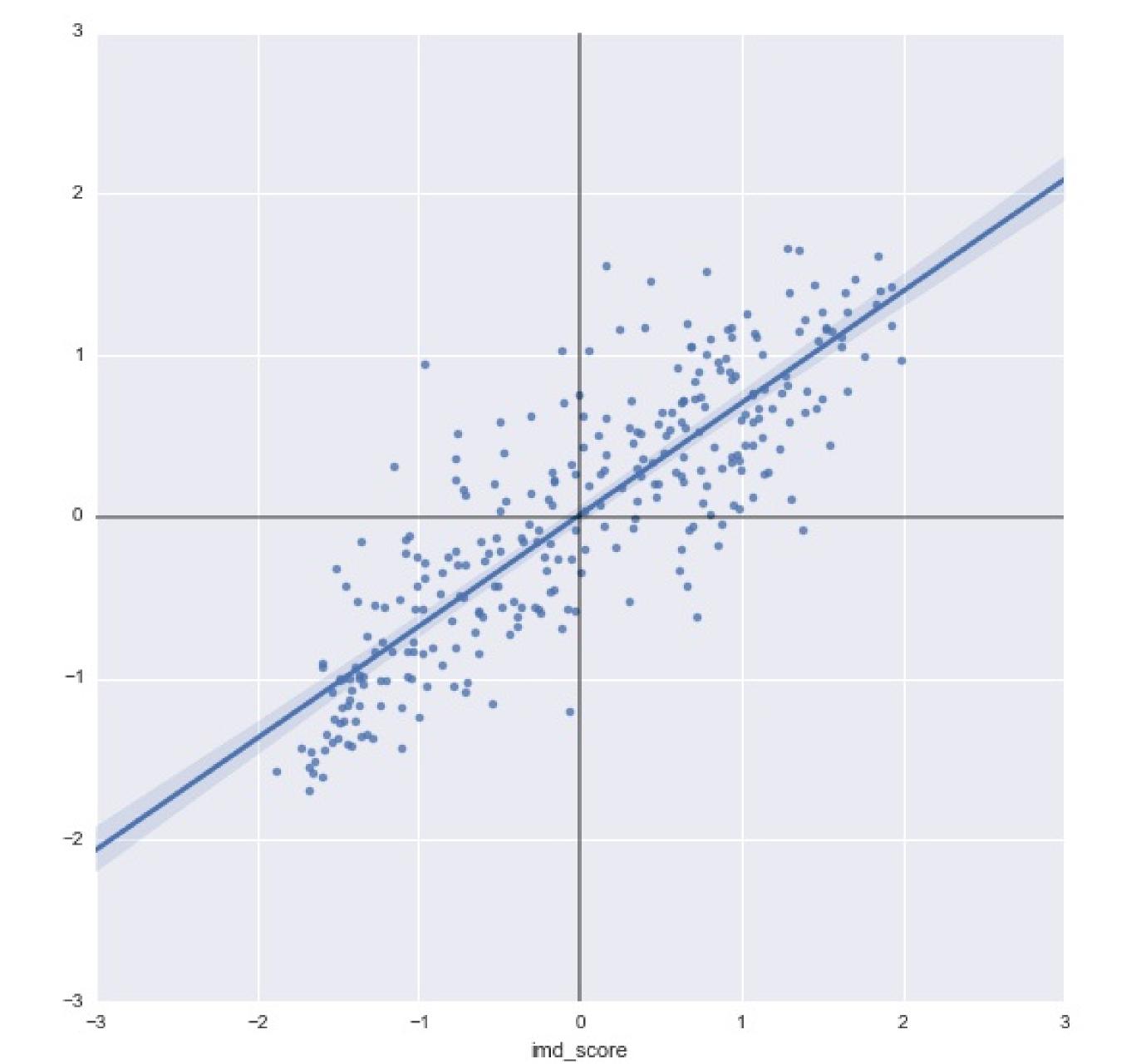


*Positive or negative correlation? Causal link?* 

- Temperature and ice-cream consumption  $\rightarrow$ **Positive.** Positive.
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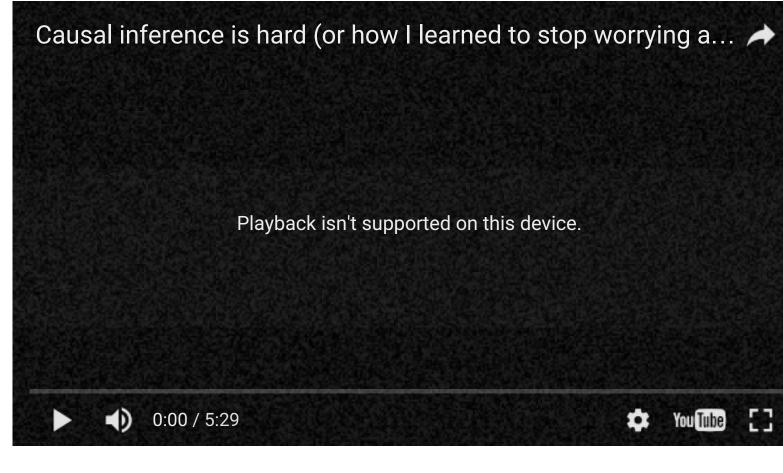
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- IMD Moran Plot in Liverpool  $\rightarrow$  **Positive**. ?

### 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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### 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

Causation *implies* Correlation

Correlation *does not* imply Causation

Why?

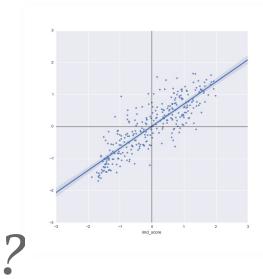
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- Reverse causality
- Confounding factors/endogeneity

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E.g. Education and income

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Two variables are correlated because they are *both* determined by other, unobserved, variables (factors) that *confound* the effect

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E.g. Ice cream and cold beverages consumption

### Strategies

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

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# StrategiesRandomized Control TrialsTreated and control groupsProbability 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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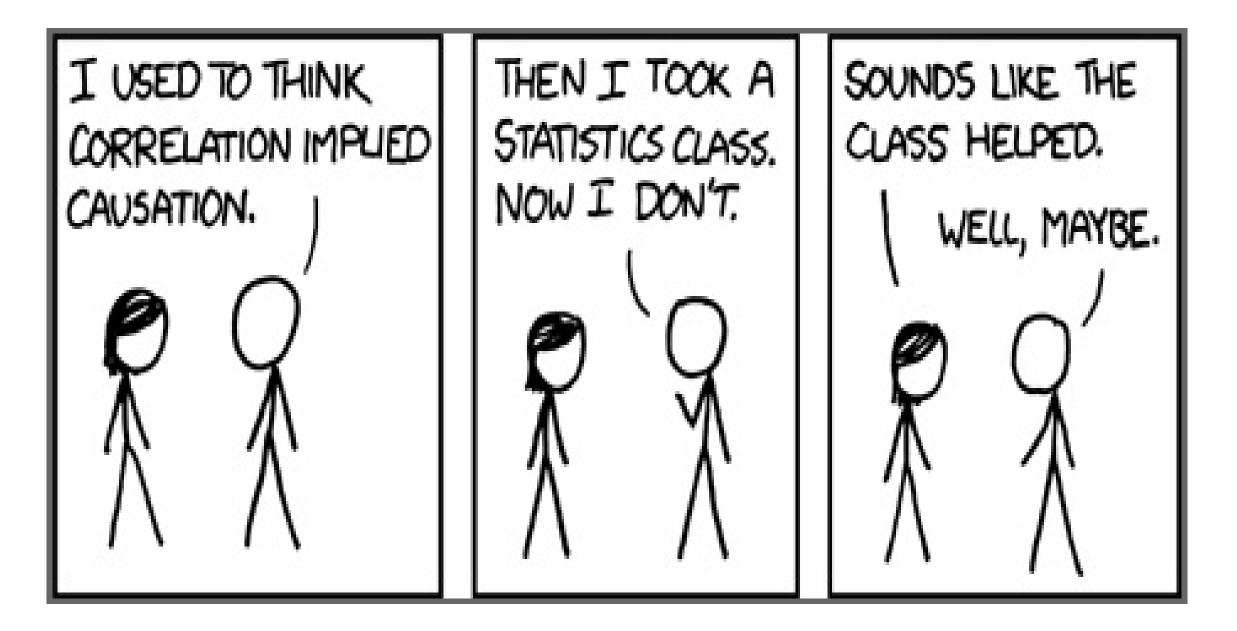
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It is important to always draw **conclusions based on analysis**, know what the data can and cannot tell, and stay **honest**.

# 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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