

Causality

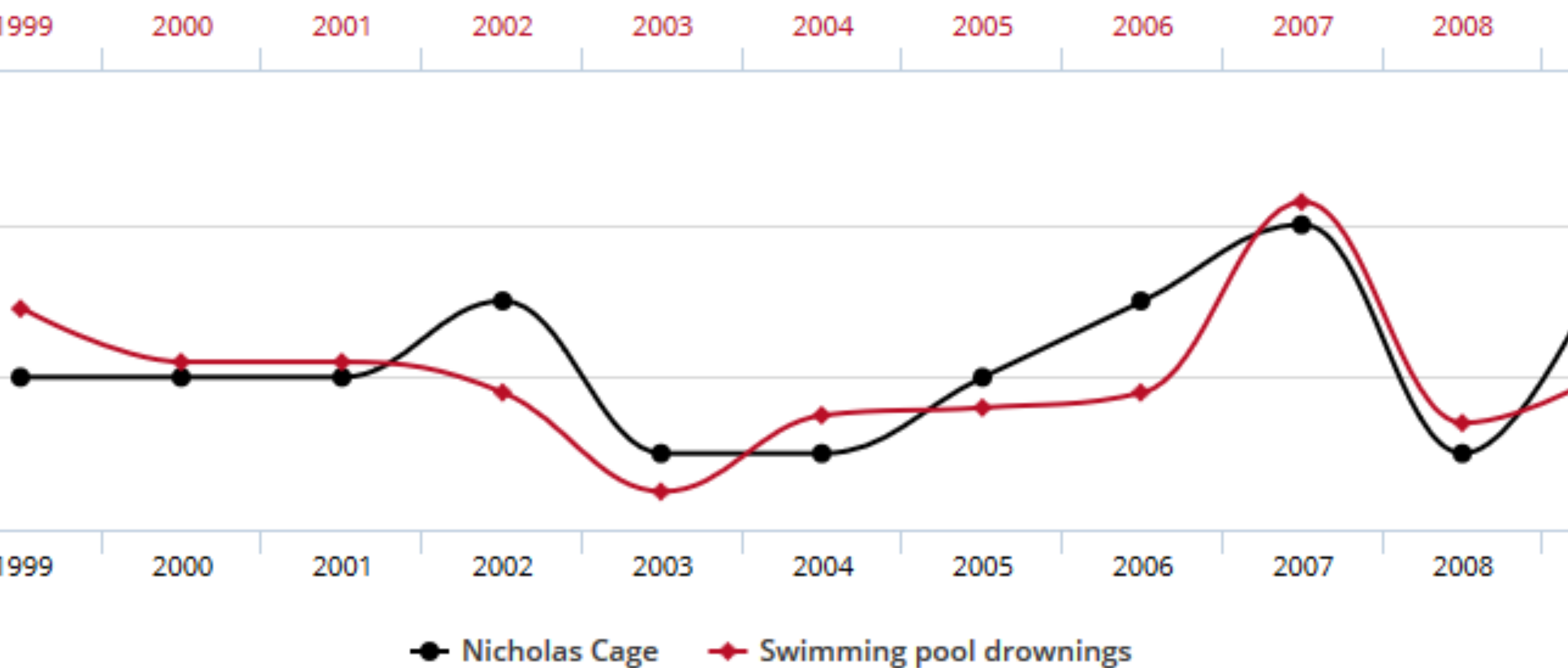
Julian Faraway



Number of people who drowned by falling into a pool correlates with

Films Nicolas Cage appeared in

Correlation: 66.6% ($r=0.666004$, $p>0.05$)



Correlation is not Causation

**Completely
Gullible**



**Utterly
Cynical**



Correlation is evidence of Causation



David Hume

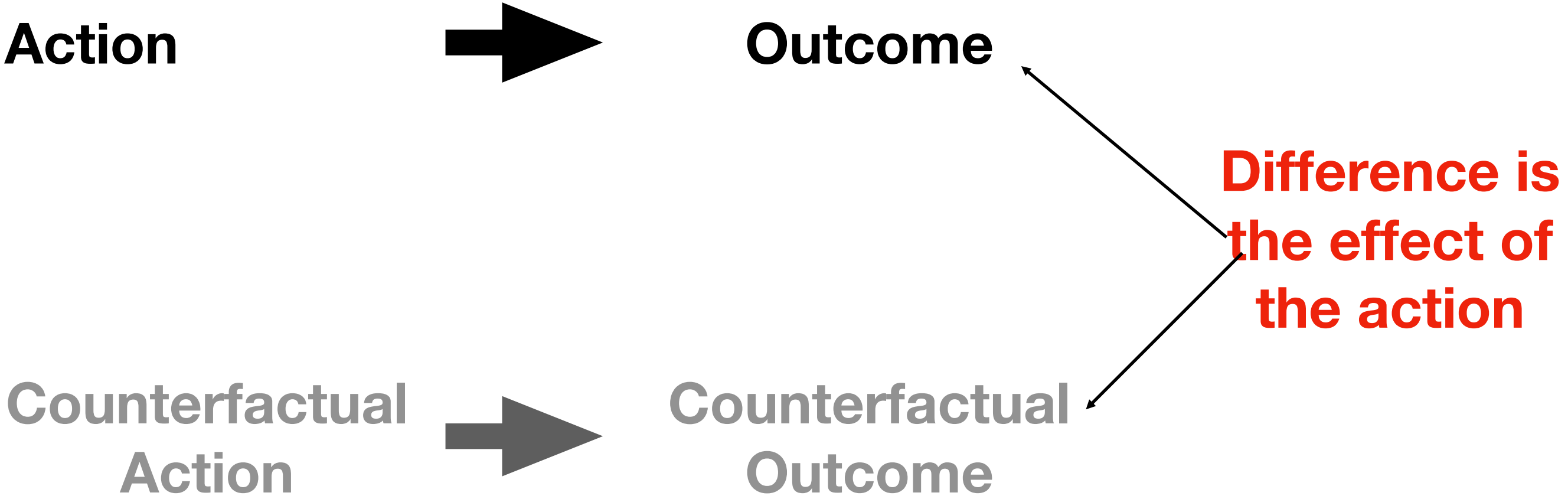
*Enquiry Concerning
Human Understanding*

1748

introduces the idea of the
counterfactual

An object is the cause of another..

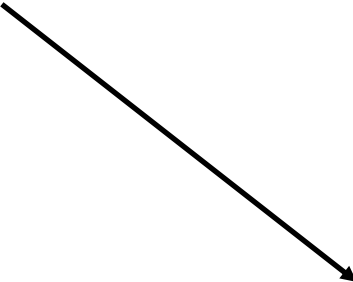
*“if the first object had not been,
the second never had existed”*



**High Air Pollution
in Ulaanbaatar**



**Poor Health
outcomes**

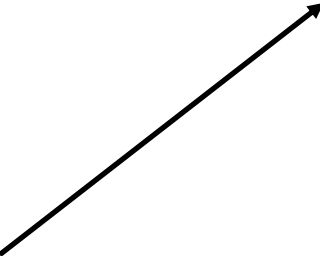


Difference?

**Lower Air Pollution
in Ulaanbaatar**

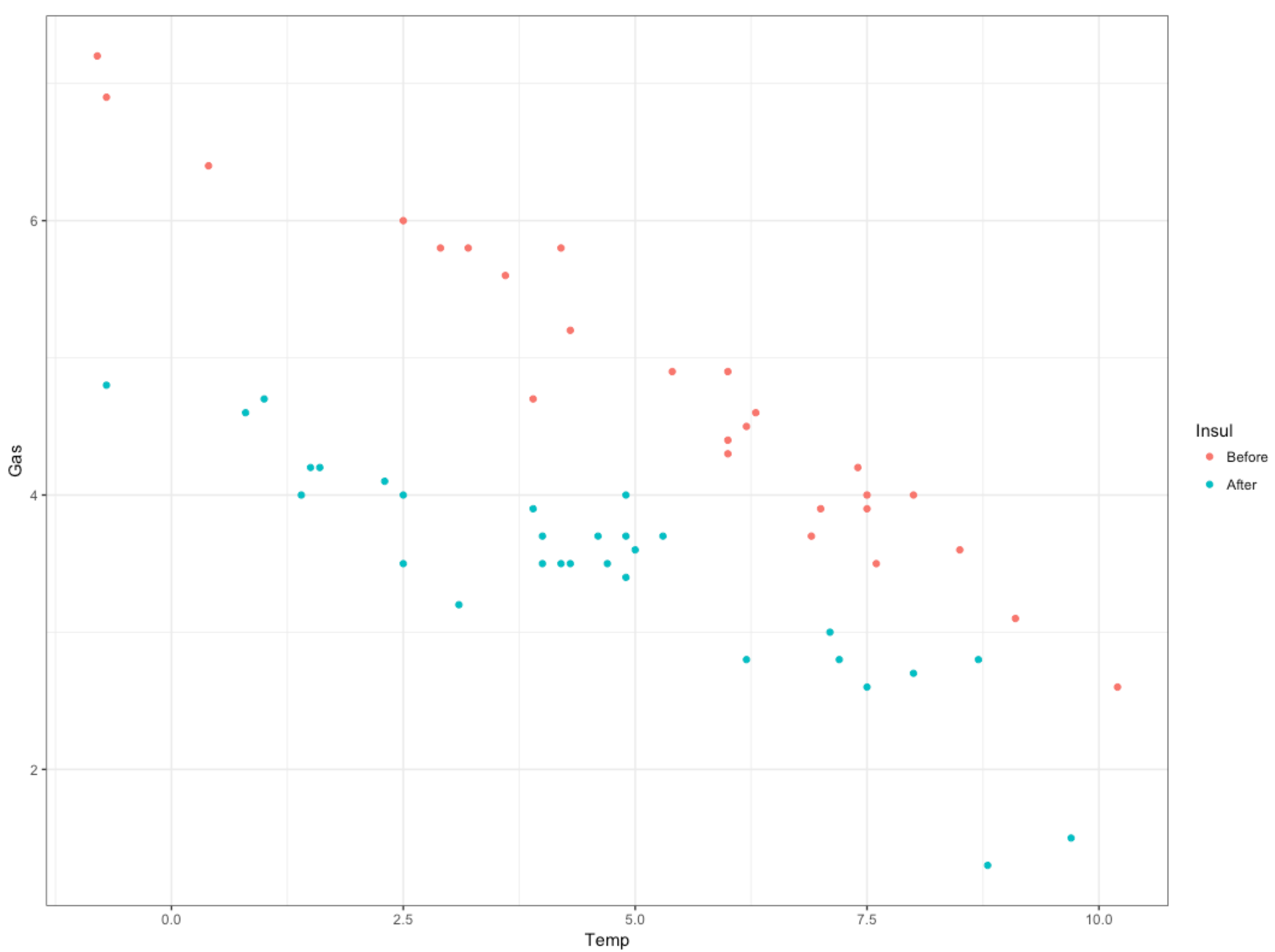


**Better Health
Outcomes?**



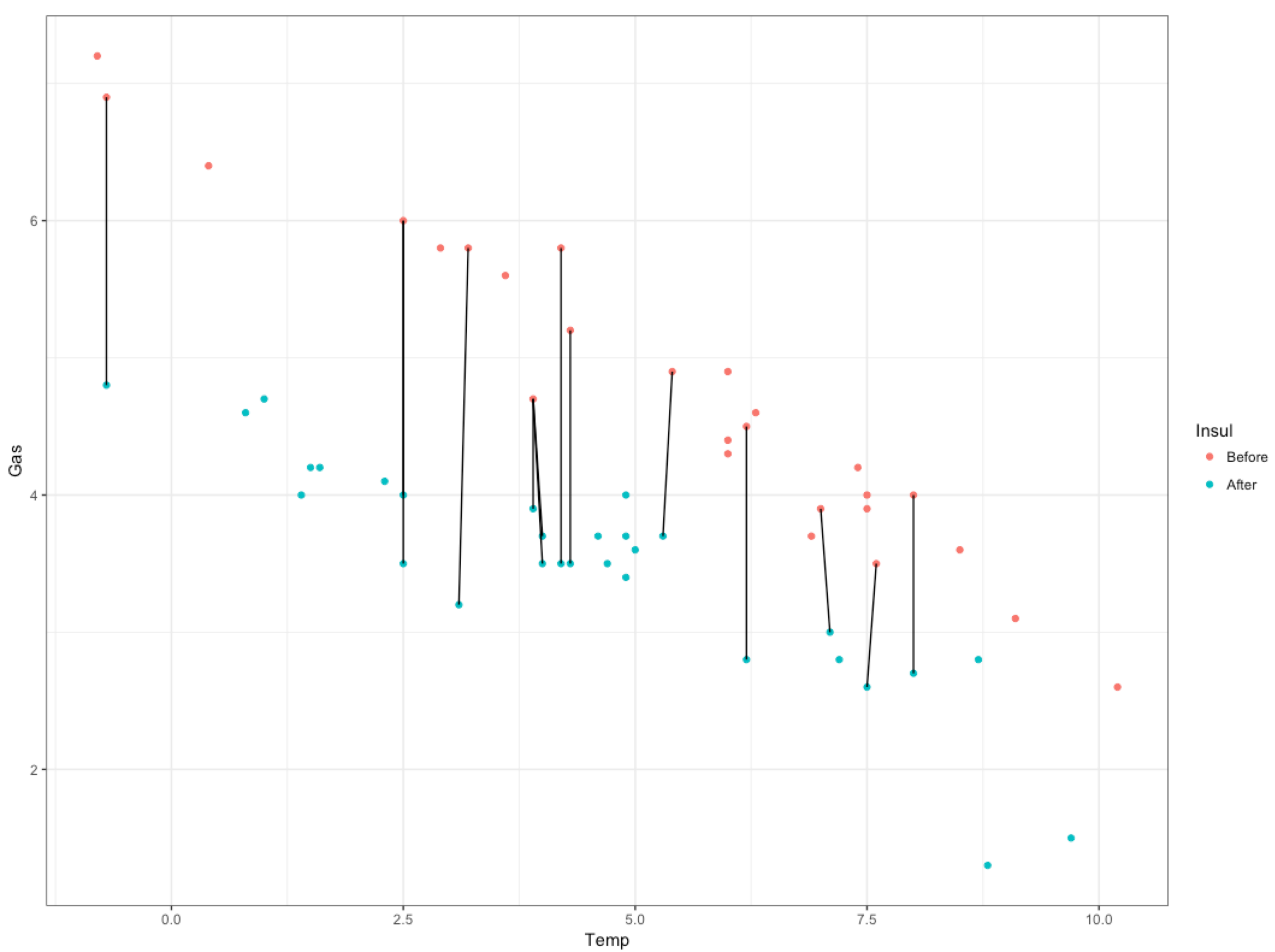
Problems with Counterfactuals

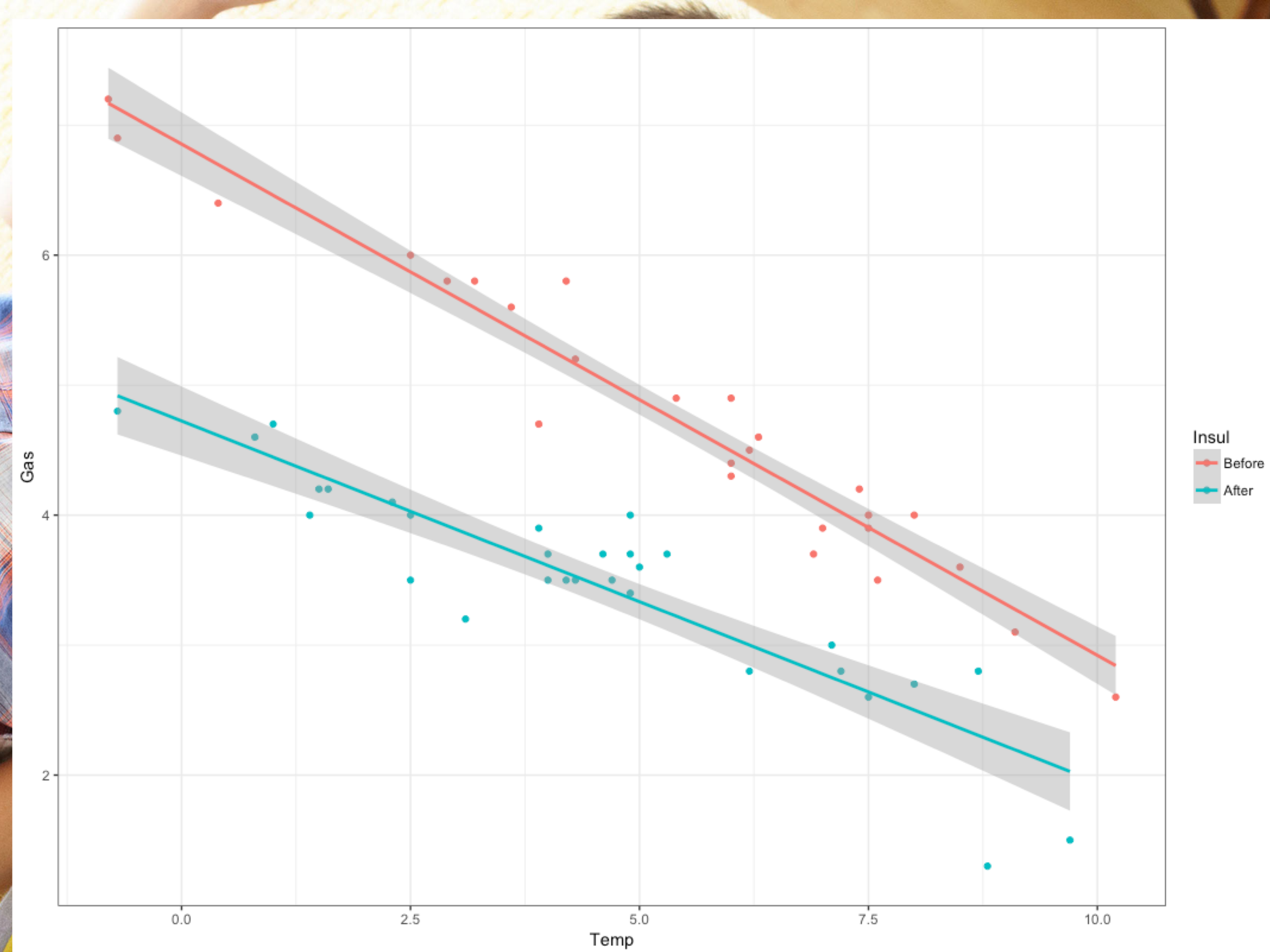
- Never get to observe
- Can the counterfactual action/decision/assignment potentially occur? What if I bought a new car instead of repairing the old one? What if I were the other gender?
- Counterfactual approximations in space and time.
 - Air Pollution/Health in Canadian Cities?
 - Air Pollution/Health in Ulaanbaatar in the past?



Matching

- Identify pairs of observations that are identical/similar in the covariates but have different treatments.
- Attempts to approximate the counterfactual outcome
- Exact matching can be difficult where observations are multivariate – *propensity score* matching can help
- Unobserved characteristics of the pair may be different in a consistent manner thus voiding the causal conclusion

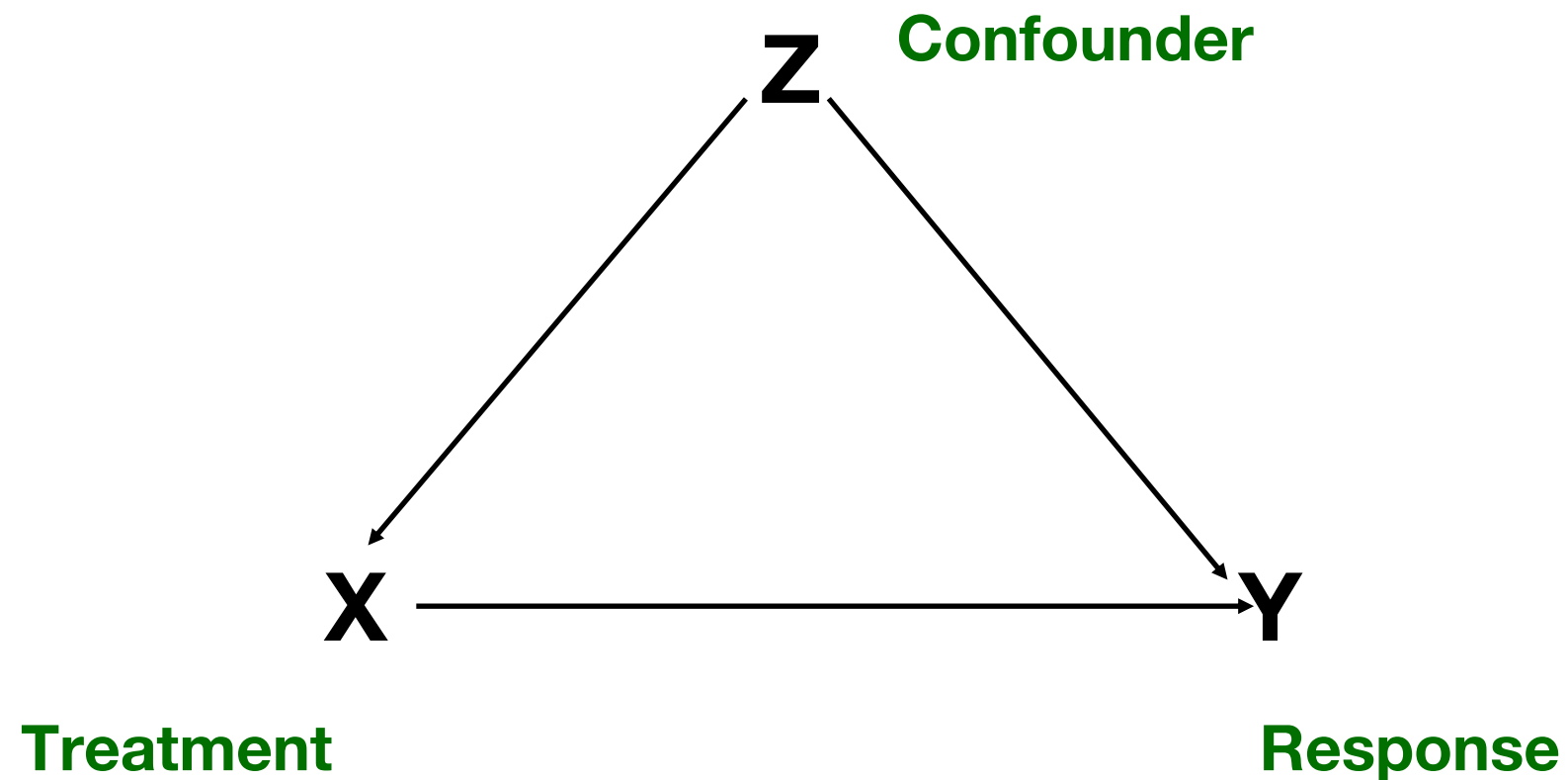




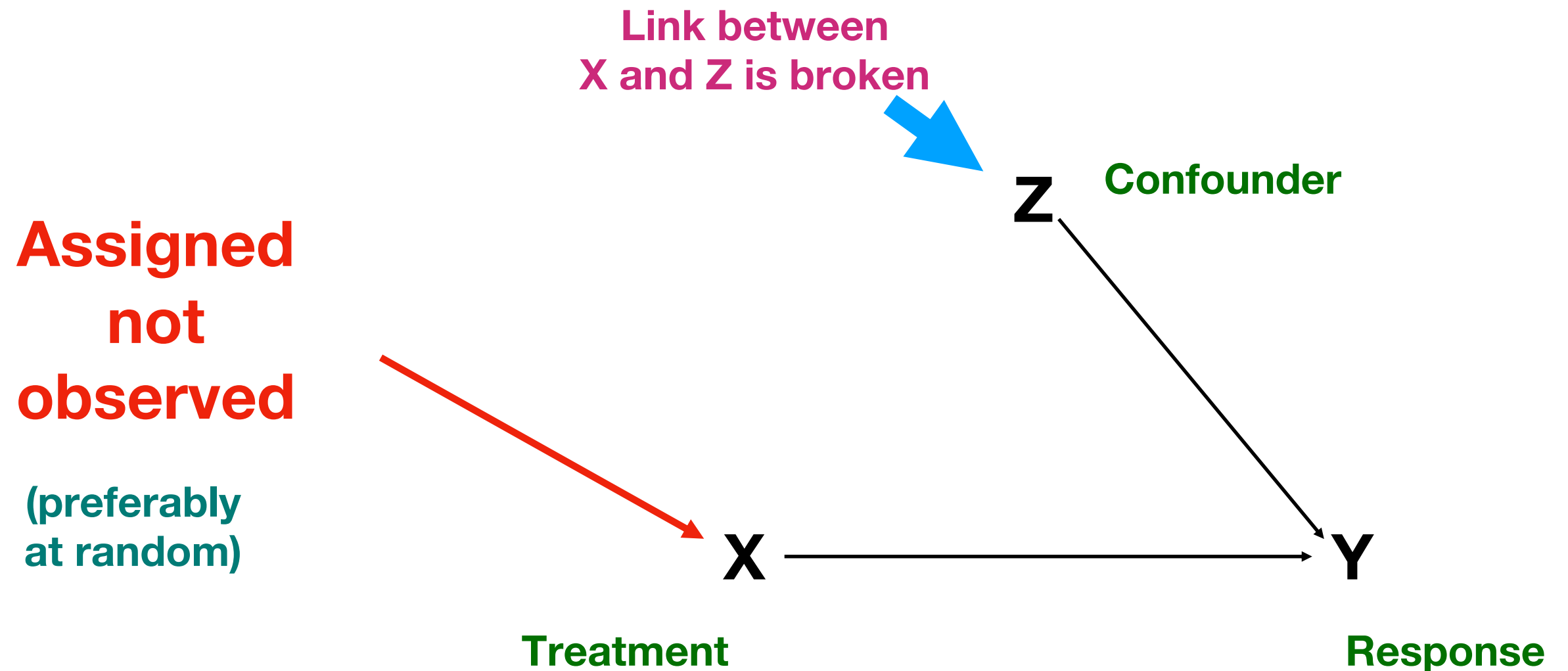
(Linear) Modelling

- Difference between the treatments is expressed as parameter(s) in the model.
- Avoids the need for matching
- Requires additional assumptions
- Non-treatment predictors are called *confounders*. Including such variables in the model is called *adjusting for the confounder*.

Graphical Models



Experiment



Judea Pearl
& Dana Mackenzie

The
Book
of
Why

The New Science
of Cause and Effect
allen lane

Statistical/Probabilistic
reasoning alone cannot
support causal inference

Determining the joint
probability distribution of
variables alone says
nothing about causation

$P(\text{Disease} \mid \text{Symptom})$

Pearl promises to determine the necessary set
of non-data assumption that are sufficient to
make a causal conclusion

Bradford Hill Criteria

Strength

Consistency

Specificity

Temporality

Biological
Gradient

Plausibility

Coherence

Experiment

Analogy

Qualitative and somewhat specific to epidemiology