

Current trends in philosophy of science

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Megatrends in PoS

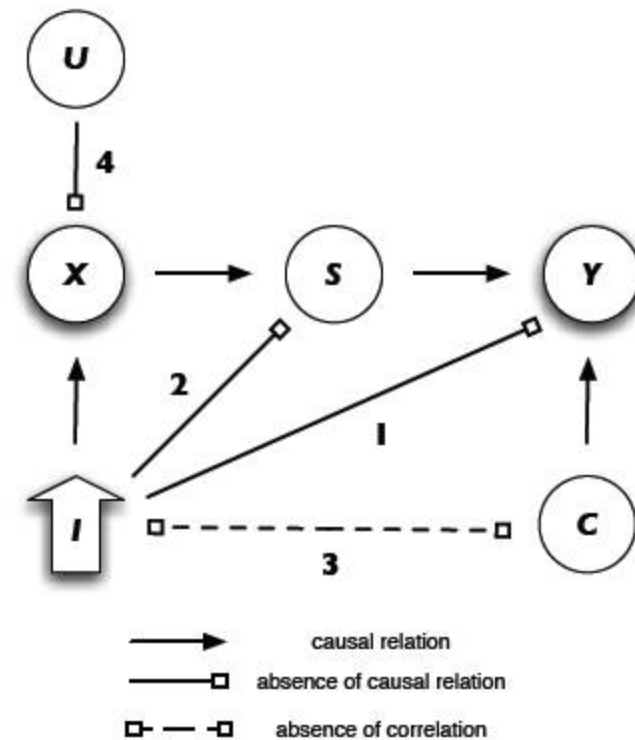
- Naturalism
 - Just say no to *a priori* philosophizing
 - Closer connection to practice
 - Argumentative resources from empirical sciences
 - Change in formal tools: from logical reconstructions to probability mathematics/statistics and computer simulation
- Philosophies of the special sciences
 - "Global" philosophical problems and "isms" less central
 - Questions closer to real methodological problems
 - → increasing specialization and fragmentation even within the philosophies of the special sciences
- Some buzzwords: model, mechanism, evidence

A solved philosophical problem: the concept of causality

- Causality and manipulability: G. H. von Wright, James Woodward, Judea Pearl
- *X causes Y, if we can bring about X by bringing about Y.*
 - This cannot be a reductive analysis of causation, because 'bringing about' is already a causal notion.
 - all proposed reductive analyses have failed, maybe we should be satisfied with a descriptive analysis
 - Goes nicely with the idea that controlled experiments are the best way to achieve causal knowledge.
- Woodward: causality as *invariance under interventions*
- a natural way to understand the difference between a *real causal relation and a mere correlation*

an ideal intervention

- *I* changes the value of *Y* only via a change in *X*:
- 1: *I* does not change *Y* directly
- 2: *I* does not change the value of some causal intermediate *S* between *X* and *Y* except by changing the value of *X*
- 3: *I* is not correlated with some other variable *C* that is a cause of *Y*
- 4: *I* acts as a switch that controls the value of *X* irrespective of *X*'s other causes *U*

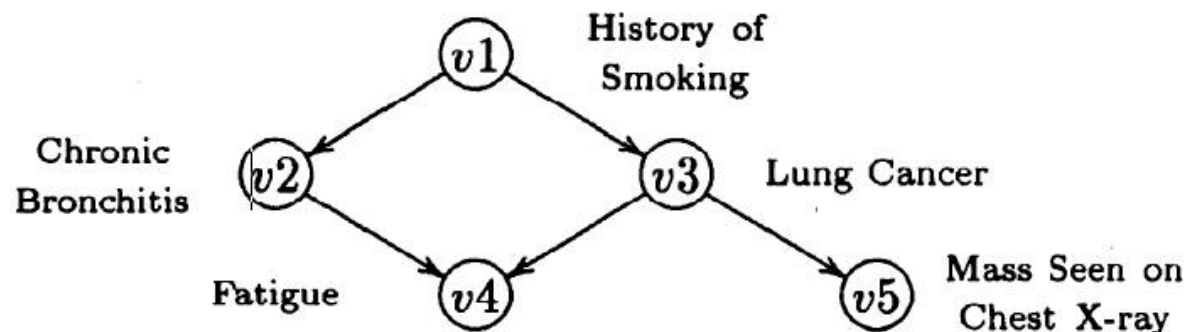


Why interventions?

- distinguish dependencies in the world from inferential dependencies
 - Observing vs. manipulating
($E[X|Y = y']$ not the same as $E[X| \text{Do}(Y = y')]$)
- define causal order (asymmetry)
- disambiguate between different causal concepts
 - (contribution, net cause, condition vs. actual cause...)
- clarify cases of confounding and multiple causal pathways
- conceptual link to manipulation
- → An interdisciplinary (philosophy, statistics, computer science), formal theory of causality

Causal reasoning and inference

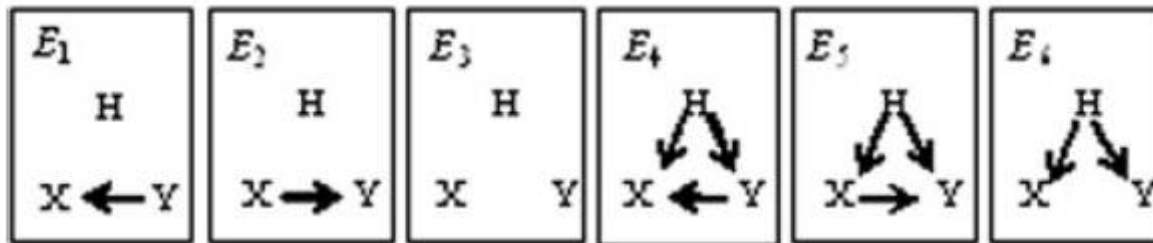
- Causal reasoning: deriving causal conclusions from a known causal structure
 - ($E[X|Y = y']$ not the same as $E[X|Do(Y = y')]$)!
 - Causal knowledge is required for predicting the consequences of exogenous interventions, regularities are enough for passive prediction.
- Causal inference: inferring the causal structure from (observational or experimental) data



Example: Causal Discovery as a Game

Eberhardt, F. (2008) Journal of Machine Learning Research 1

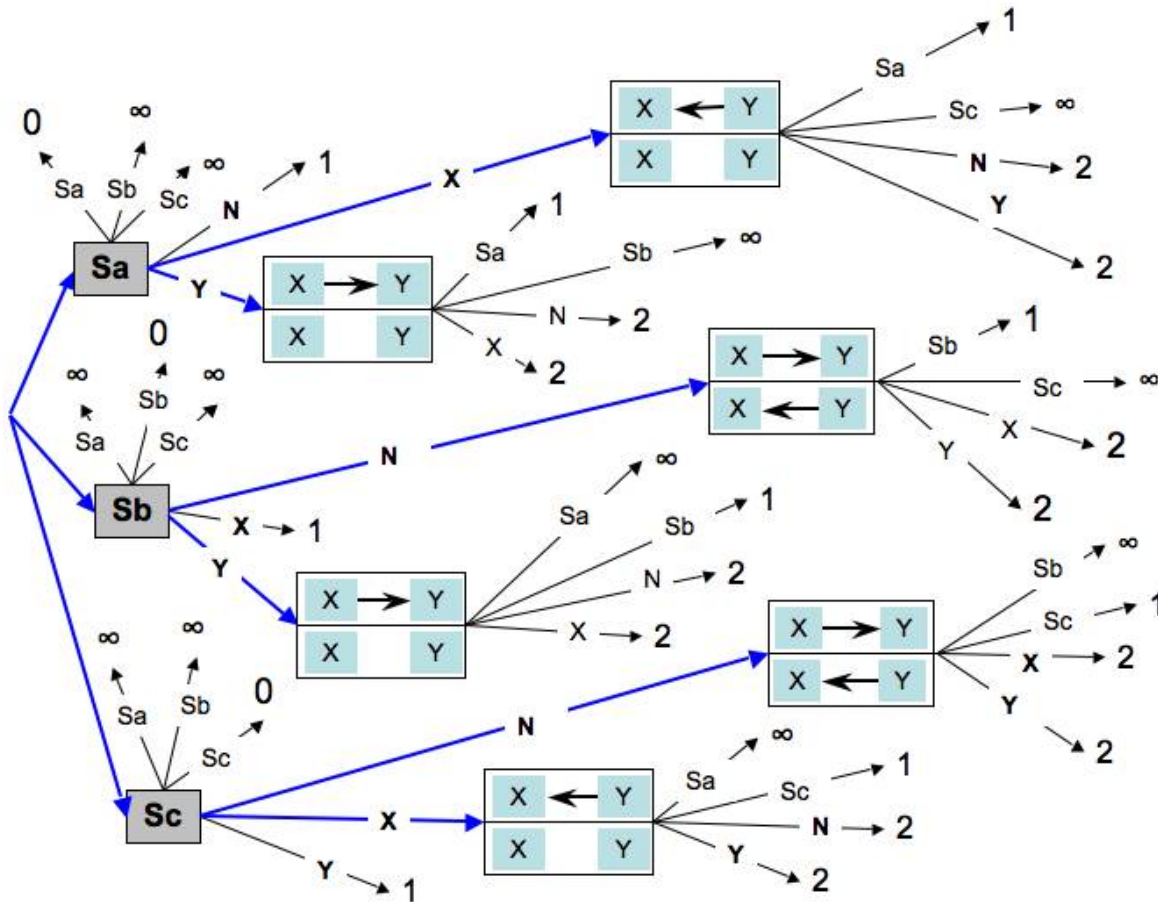
- The problem: How to select the optimal experiments?
- Intervention = randomized controlled trial
- Causal discovery as a (zero-sum) game between players *Scientist* and *Nature*.
 - Scientist's strategies: what (passive) observations or RCTs to perform and when to make a guess about the structure
 - Nature's strategies: select the true structure



Example: Causal Discovery as a Game

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- $Sa := X \rightarrow Y$; $Sb := X \rightarrow Y$ and $Sc := X \leftarrow Y$:

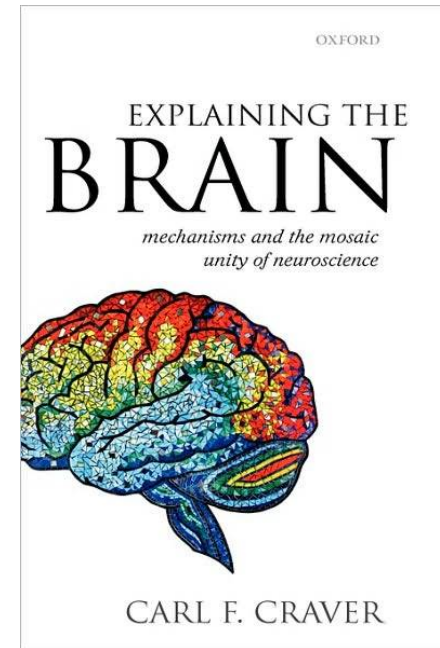


From laws of nature to mechanisms

- Traditional nomothetic picture: sciences aim at uncovering universal, exceptionless regularities.
- Mechanism: a structure performing a function in virtue of its component parts, component operations, and their organization. The orchestrated functioning of the mechanism is responsible for one or more phenomena.
- Causation and explanation: explanatory relationships not "natural laws", but *invariances* realized by mechanisms
- Confirmation and evidence: from inductive logics to causal discovery algorithms, reductivist heuristics and mechanistic extrapolation
- Conceptualizing the unity of science: not deduction of special laws from more fundamental laws (classical reduction), but relating *levels of mechanisms*

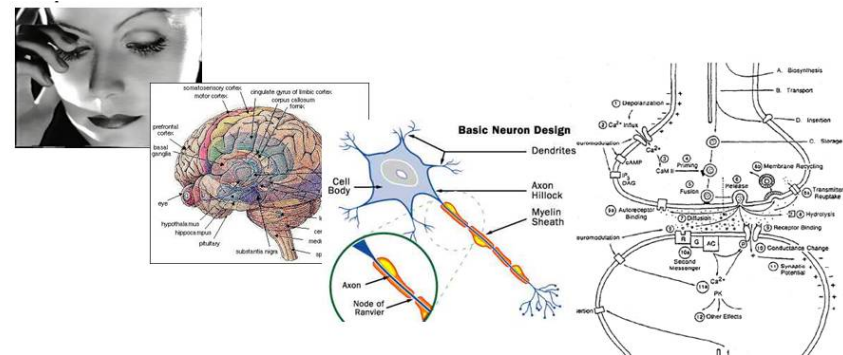
Mechanistic explanation

- Carl Craver: Explaining the Brain
 - Explanations describe mechanisms
 - Explanations are multi-level
 - Levels of mechanism
 - Levels are local, behaviour → mechanism → components
 - Intra-level, as well as inter-level integration of fields
 - research not always bottom-up or top-down
 - Evidential relevance as constraining the space of *plausible mechanisms*
 - Manipulability as the criterion of explanatory relevance
 - *Mechanisms schemas* as the blueprint for the integration of scientific knowledge



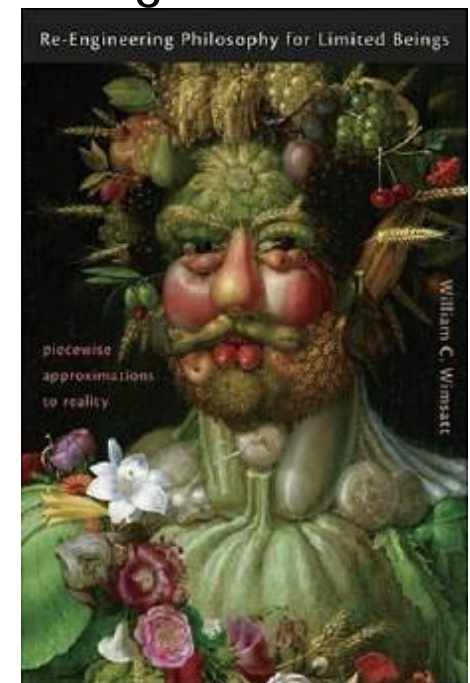
Mechanistic research programmes and reuctionistic heuristics

- Opening black boxes according to the heuristics of *functional decomposition and localization* (Bechtel and Richardson 2000)
 - phenomena that the system of interest exhibits are identified
 - phenomenon of interest is *functionally decomposed*
 - the system is *structurally decomposed*
 - *localization* of the component operations to appropriate structural component parts



Mechanistic research programmes and reductionistic heuristics

- William Wimsatt: The point of “reductionism”: make simplifying hypotheses in order to learn about the organization by finding out why simple hypotheses *fail*.
- Reductionistic biases:
 - Localization fallacies in conceptualization, modeling and testing
 - Functional localization fallacies
 - Interface determinism
 - ...



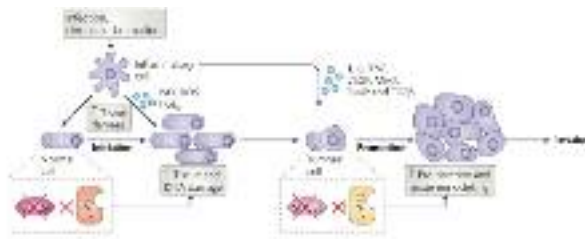
Emergence as a failure of aggregativity

- William Wimsatt: a property of a system is *aggregative with respect to* a decomposition to its parts and their properties, when it is invariant with respect to
 - *Intersubstitution*: rearranging parts or interchanging parts with (relevantly) equivalent ones
 - *Size scaling*: addition or subtraction of parts
 - *Decomposition and reaggregation*: decomposition and reaggregation of parts
 - *Linearity*: there are no cooperative or inhibitory interactions between the parts
- *Don't ask whether it is emergent, but how it is emergent!*
 - most things outside fundamental physics (properties subject to conservation laws) are emergent
 - learning about organization by exploring the limits of aggregation

Example: mechanistic extrapolation

Steel, Daniel (2008): *Across the Boundaries: Extrapolation in Biology and Social Science*, OUP

- The problem: *the extrapolator's circle*: how is it possible to establish the similarity of the model and target without already knowing what one wants to extrapolate?
- How to reduce the number of required comparisons:
 - background knowledge according to which causally relevant disanalogies are likely to be found at some stages of the mechanism and not others.
 - comparisons of model and target mechanisms will be more efficient if they focus on mechanism activities and components that are *downstream in the sense of being more direct causes of the outcome*
→ “*distinctive markers*”



Literature

- Bechtel, William and Richard Scheines 2000: *Discovering Complexity: Decomposition and Localization as Strategies in Scientific Research (2nd ed.)*, Princeton NJ: Princeton UP.
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