

# Foundations in Causal Thinking for Health Data Statisticians

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On behalf of TG7

everything that becomes or changes must do so owing to some cause; for nothing can come to be without a cause

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*Plato*

- “Epidemiological research is, almost exclusively, concerned with *etiology* of illness”, where etiology = causal origin of illness (Miettinen & Karp, 2012).
- In fact, the goal of most statistical analyses is to uncover causal relationships.

- There is no agreement on the definition of **causality**, particularly across disciplines (or across centuries!).
- In 1890, Koch proposed criteria to establish a ‘causative relationship’ between a microbe and a disease; imperfect but reasonable - but only for pathogens.
- Pearl (2009, p. 25-26), a computer scientist and a leader in the field of modern causality, does not explicitly define causality at all, refers to causal relationships as “stable” and “ontological”.
- Meinshausen, Peters & Buehlmann (2016) similarly deem causal relationships to be present when multiple data sources produce ‘invariant prediction’.

- Earliest, and best known, ideas in epidemiology on causality are the (non-)criteria given by Sir Austin Bradford-Hill in 1965:
  1. Strength
  2. Consistency
  3. Specificity
  4. Temporality
  5. Biological gradient
  6. Plausibility
  7. Coherence
  8. Experiment
  9. Analogy
- A group of conditions to *assess* (not establish) causality.

- Less well-known is Bradford Hill's wide-ranging lecture on 'The Statistician in Medicine', recently reprinted in *Statistics in Medicine* in celebration of 40 years since its inception.
- Three themes:
  - ▶ knowledge of the area of application,
  - ▶ types of data to provide evidence and how it is gathered (including the poor experiment that is 'nature'),
  - ▶ drawing conclusions from evidence.
- Still very relevant!

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- Bilingualism!
- Context is critical implementing any analysis
- Subject-matter knowledge must be used to inform modelling, but equally critical to note what is known vs. what is assumed, and whether the data themselves were collected fairly.
- DAG

# ‘The Statistician in Medicine’: On medicine

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N. Röhrig et al. / Journal of Clinical Epidemiology 67 (2014) 199–206

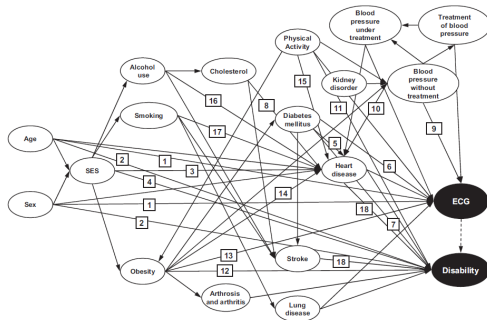


Fig. 1. DAG derived from literature and expert knowledge. Nodes represent variables and arrows represent causal associations. Dark-colored nodes label ECG findings and disability, representing exposure and outcome, respectively. Pale-colored nodes represent possible confounding factors. Numbers represent available information from the literature (see Table 1 at [www.jclinepi.com](http://www.jclinepi.com) for full references). SES, socioeconomic status; ECG, electrocardiography.



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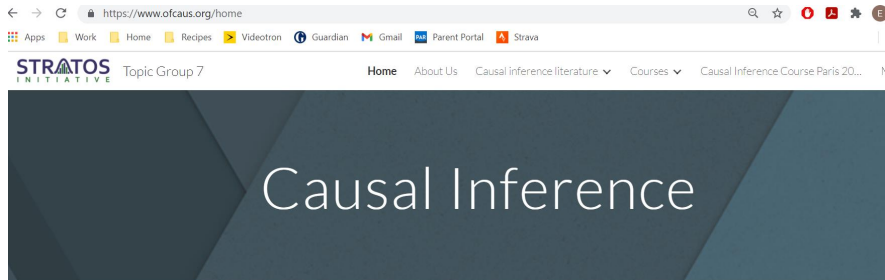
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- ...so correlation *may be* causation. (Just be careful!)

We are continuously brought back to the fundamental question – what alternative explanation will fit a set of observations, what other differences between our contrasted groups could equally, or better, account for the observed incidences.

- This is where the ‘causal’ methods come in: we seek methods that reduce or eliminate alternative explanations such as *imbalance* in covariates between treatment groups (confounding, selection, missing data, etc.)

# The STRATOS in Medicine: Causal inference!



The screenshot shows a web browser at the URL <https://www.ofcaus.org/home>. The page header includes the STRATOS INITIATIVE logo and 'Topic Group 7'. Navigation links are provided for 'Home', 'About Us', 'Causal inference literature', 'Courses', and 'Causal Inference Course Paris 20...'. The main content area features a large dark blue banner with the text 'Causal Inference' in white.



## STRATOS Initiative

Topic group 7 is a member of the [STRATOS Initiative](#) (STRengthening Analytical Thinking for Observational Studies) which is a large collaboration of experts in many different areas of biostatistical research. Ongoing research, discussions and activities within STRATOS are conducted in [nine topic groups](#) and several cross-cutting [panels](#).



## Formulating causal questions and principled statistical answers

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Inference (TG7) of the STRATOS initiative

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Although review papers on causal inference methods are now available, there is a lack of introductory overviews on *what* they can render and on the guiding criteria for choosing one particular method. This tutorial gives an overview in situations where an exposure of interest is set at a chosen baseline (“point exposure”) and the target outcome arises at a later time point. We first phrase relevant causal questions and make a case for being specific about the possible exposure levels involved and the populations for which the question is relevant. Using the potential outcomes framework, we describe principled definitions









# The STRATOS in Medicine: Causal inference!

## Tutorial: Formulating causal questions and principled statistical answers

On this page we include a link to our tutorial: [Formulating causal questions and principled statistical answers](#), along with a data set and accompanying code. The purpose of the data is to illustrate concepts and estimation approaches by simulating a case study inspired by the Promotion of Breastfeeding Intervention Trial (PROBIT) a large randomized study in which mother-infants pairs across 31 Belarusian maternity hospitals were assigned either standard care or the possibility to follow a breastfeeding encouragement programme (Kramer MS et al. 2001). The aim was to investigate the effect of the programme and breastfeeding on a child's later development. In our simulation we go beyond generating the 'observed data' by also simulating for every unit in the study how different exposures would lead to different potential outcomes. The data set is called the simulation learner PROBITsim.

### References

Kramer MS, Chalmers B, Hodnett ED, et al. Promotion of breastfeeding intervention trial (PROBIT) - A randomized trial in the Republic of Belarus. *Journal of the American Medical Association*. 2001;285(4):413-420.

TITLE	LAST MODIFIED
 Analysis with SAS A2.sas	5/16/19 Ingeborg Waernbaum
 Analysis with SAS A3_A0.sas	5/16/19 Ingeborg Waernbaum
 Analysis with SAS A3_A1.sas	5/16/19 Ingeborg Waernbaum
 Analysis_code.R	5/16/19 Ingeborg Waernbaum
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 Generate_simulation_learner.R	5/16/19 Ingeborg Waernbaum
 PROBITsim2018_v12.dta	11/17/19 Ingeborg Waernbaum

# The STRATOS in Medicine: Causal inference!

Tutorial: Formulating causal questions and principled  
statistical answers

...for survival outcomes (in progress)

...for time-varying exposures (future work)

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- Causal estimation can be accomplished *by design*, but often requires the additional help of *modelling* to account for unplanned imbalances.
- The statistical framework for causal inference tries to formalize assumptions and make them as clear and explicit as possible; we mustn't forget that they are present and the foundation for our conclusions.
- The stakes in medicine and public policy are high – it is worth investing energy and care to ensure our heap of evidence is convincingly high.