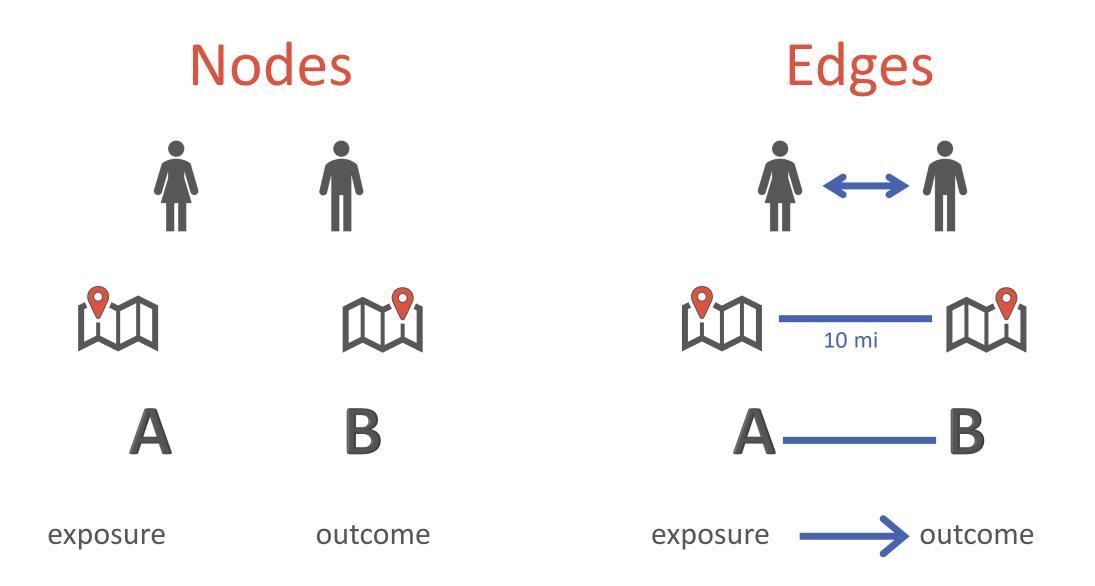
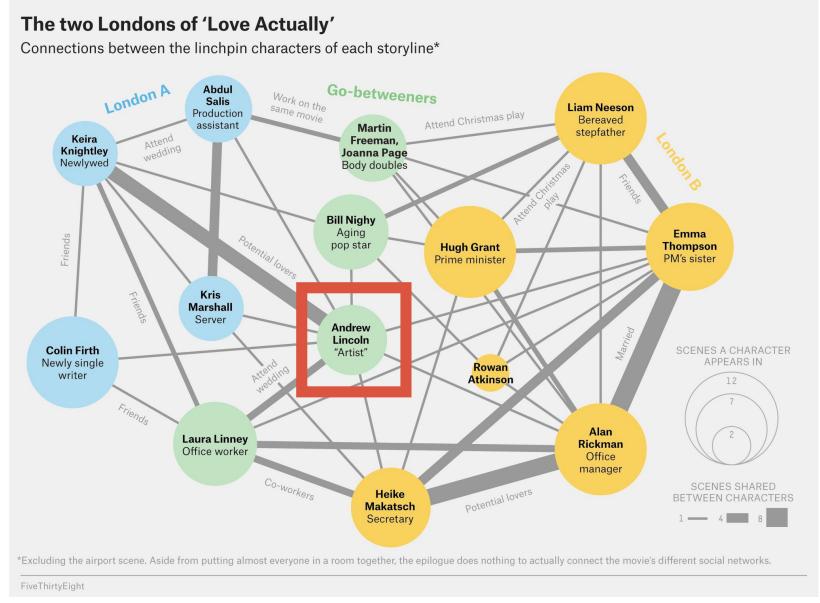
Directed Acyclic Graphs

an introduction

Louisa H. Smith Kolokotrones Symposium December 7, 2018 **Components of graphs**

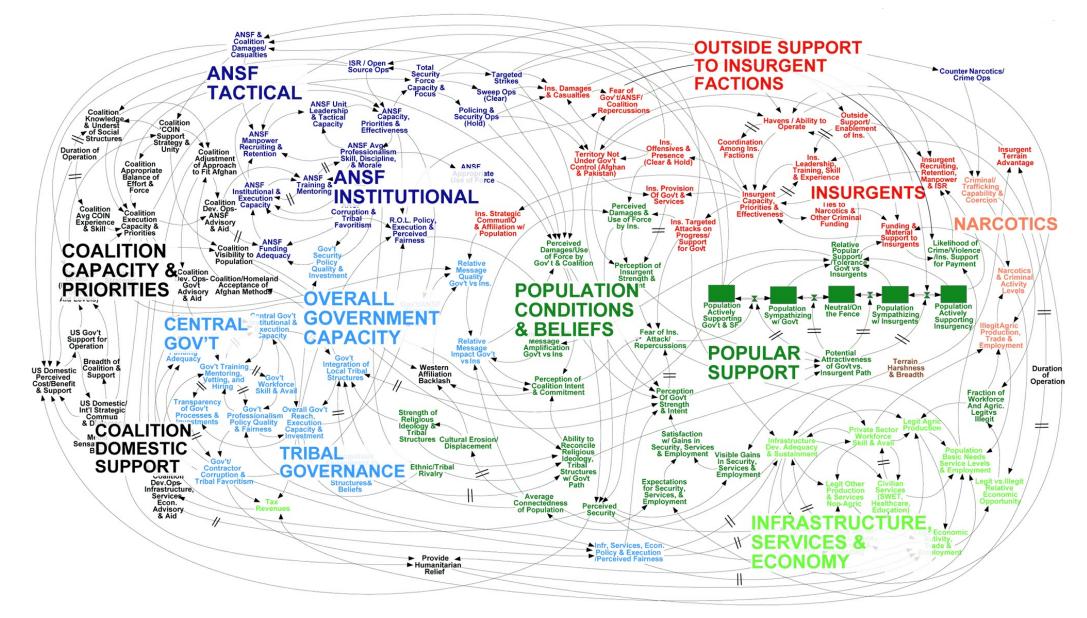


Love Actually graph



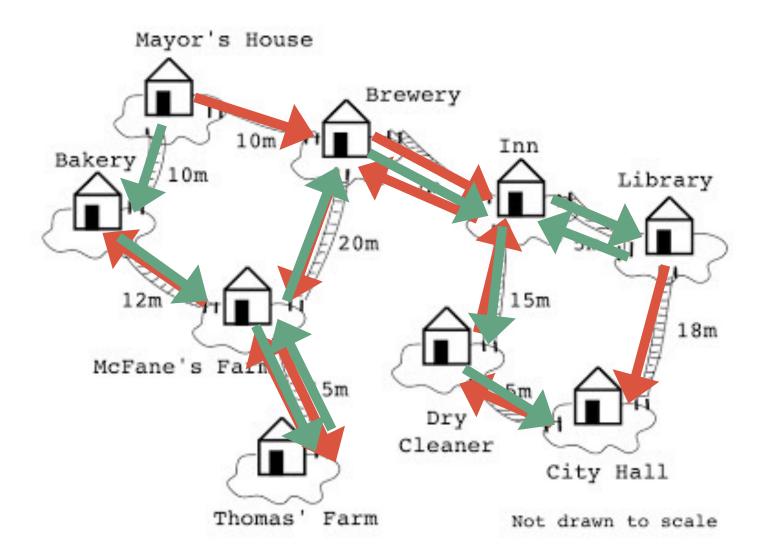
https://fivethirtyeight.com/features/the-definitive-analysis-of-love-actually-the-greatest-christmas-movie-of-our-time/

Afghanistan strategy graph



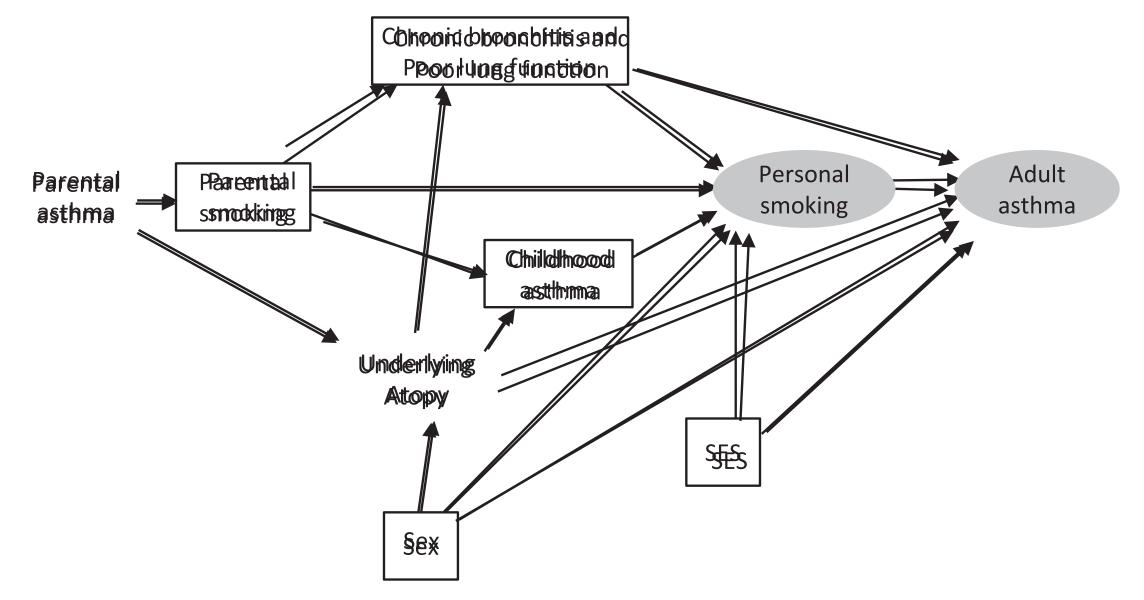
http://sdwise.com/2013/07/hey-new-york-times-a-causal-loop-diagram-is-not-a-powerpoint-fail/

Travelling salesman graph



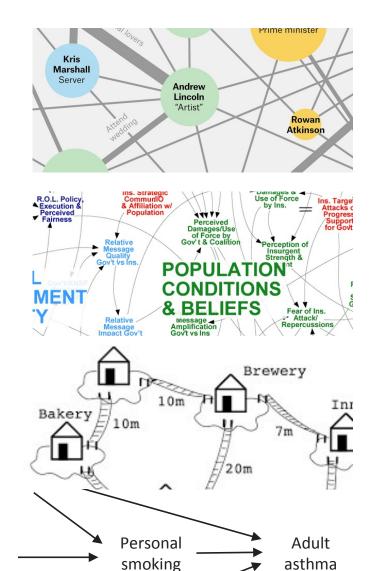
http://computationaltales.blogspot.com/2011/07/dijkstras-algorithm-on-scooters-part-4.html

Smoking-asthma graph



Williamson EJ, et al. Introduction to causal diagrams for confounder selection. Respirology. 2014;19(3):303-311.

Graphs help solve problems

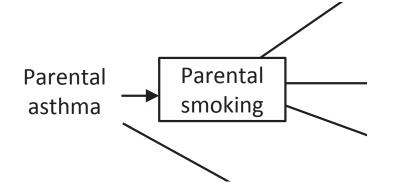


- Who is the central character tying together the various plots in the movie?
- What are the downstream effects of increasing resources in one sector vs. another?
- What is the shortest route connecting all the locations?
- How can we reduce (and avoid inducing) bias in our study?

Strategies that may reduce (or induce!) bias

Restrict a study's participants to those with certain values of a variable

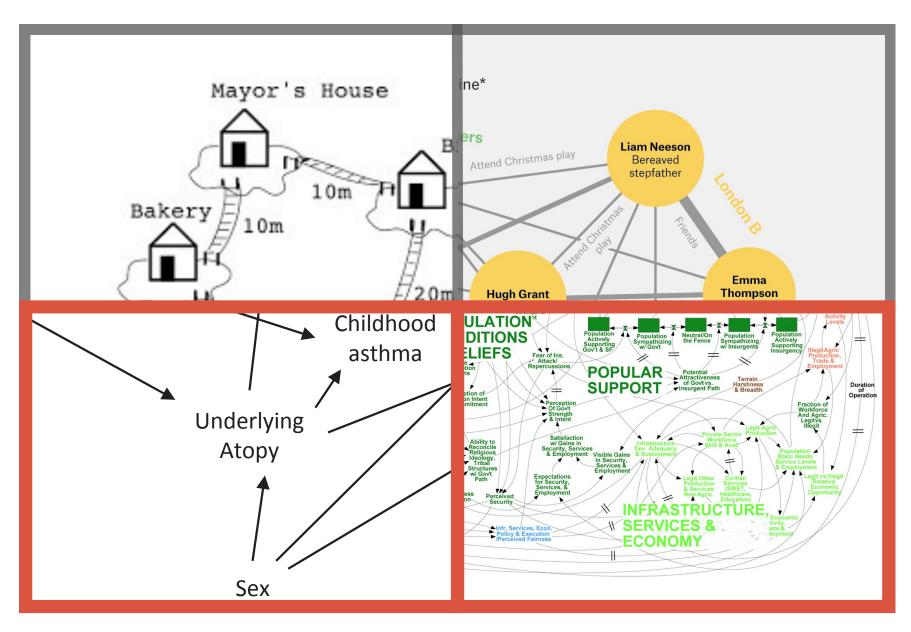
- e.g., recruit only people whose parents smoked
- Stratify by or statistically adjust for a variable
 - e.g., include parental smoking status (yes/no) in the model



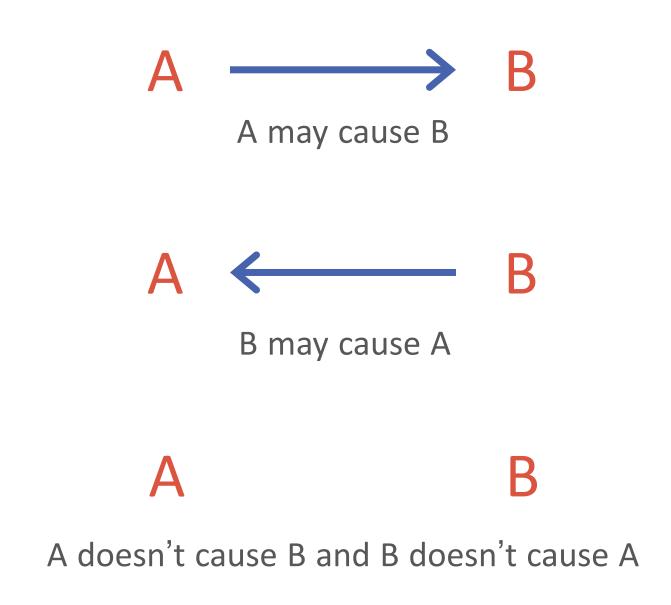
We often indicate that we've conditioned on a variable (by restricting, adjusting, etc.) by drawing a box around that node

Directed Acyclic Graphs

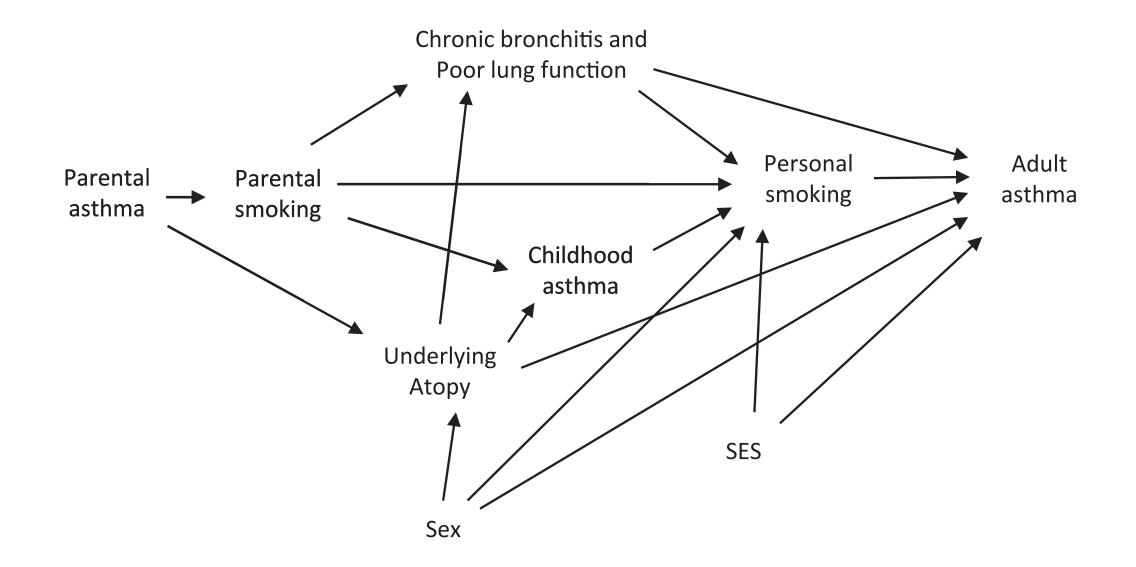
Graph edges



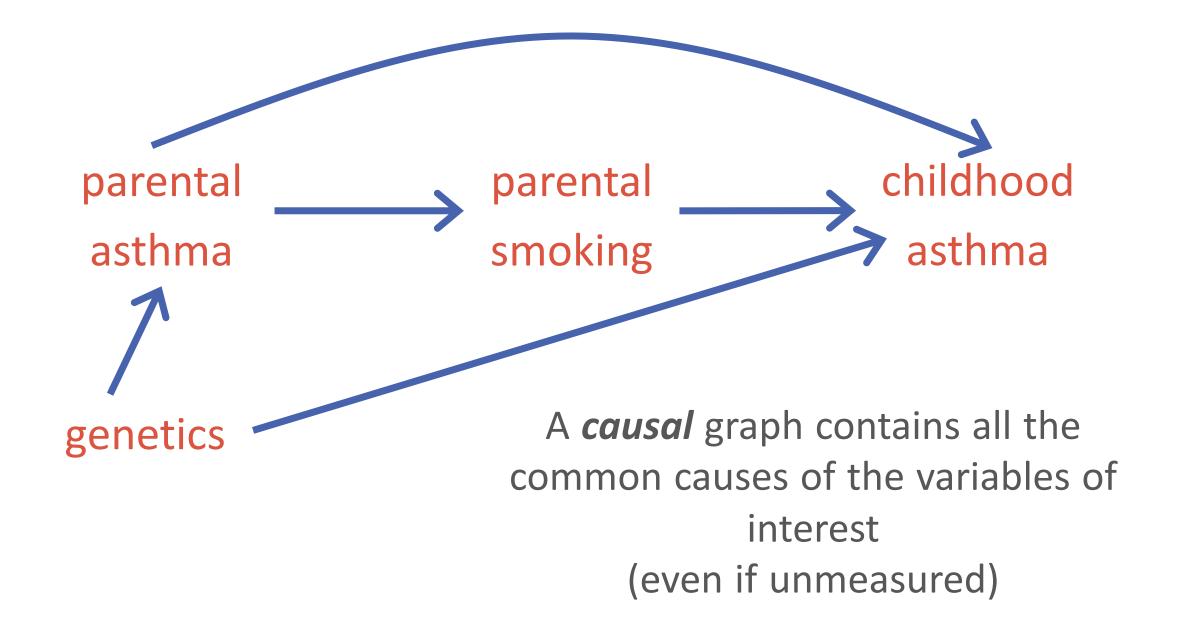
Edges are *directed*



Smoking-asthma relationship assumptions

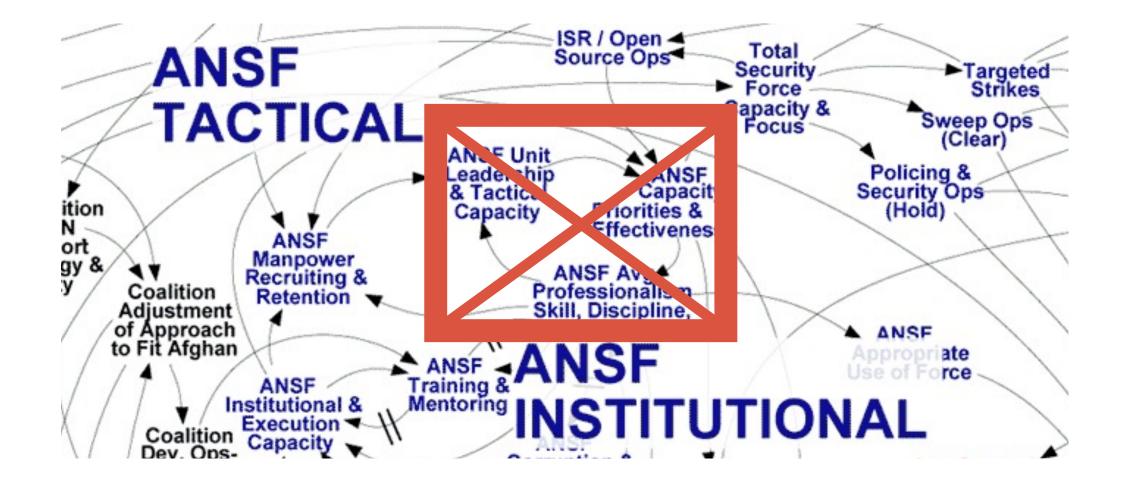


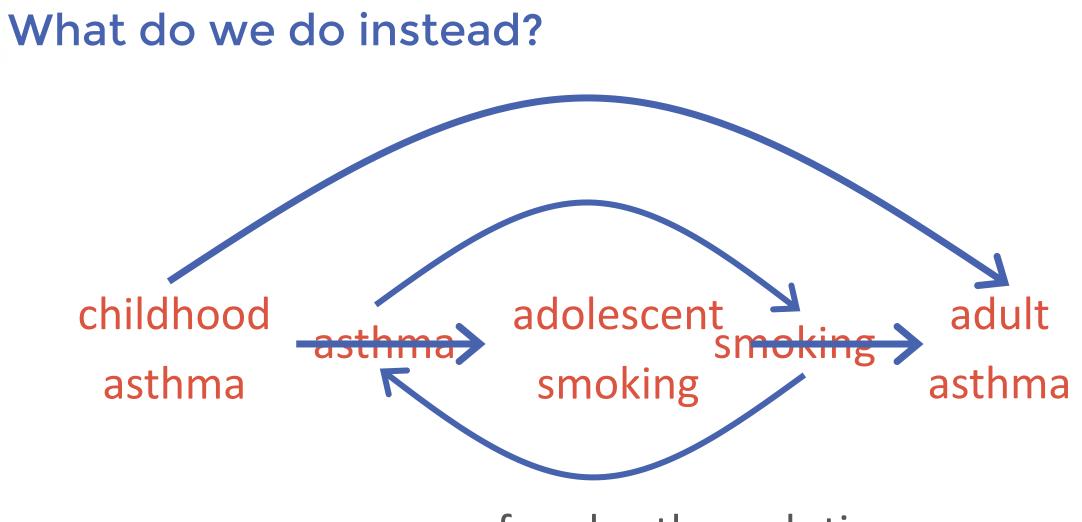
Smoking-asthma *causal* relationships



Directed Acyclic Graphs

Feedback loops

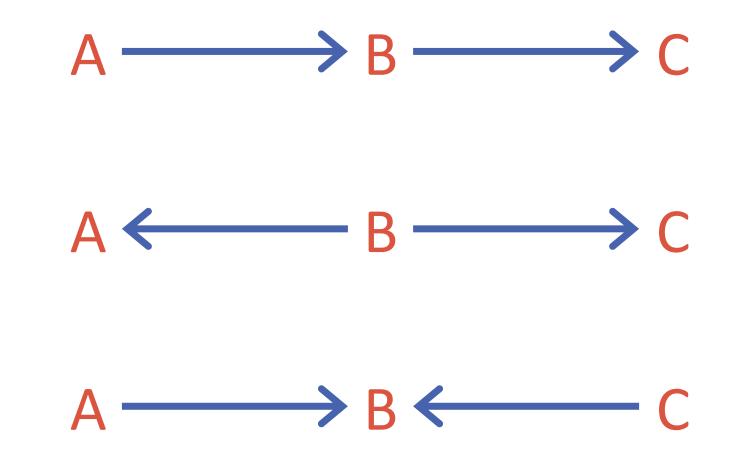




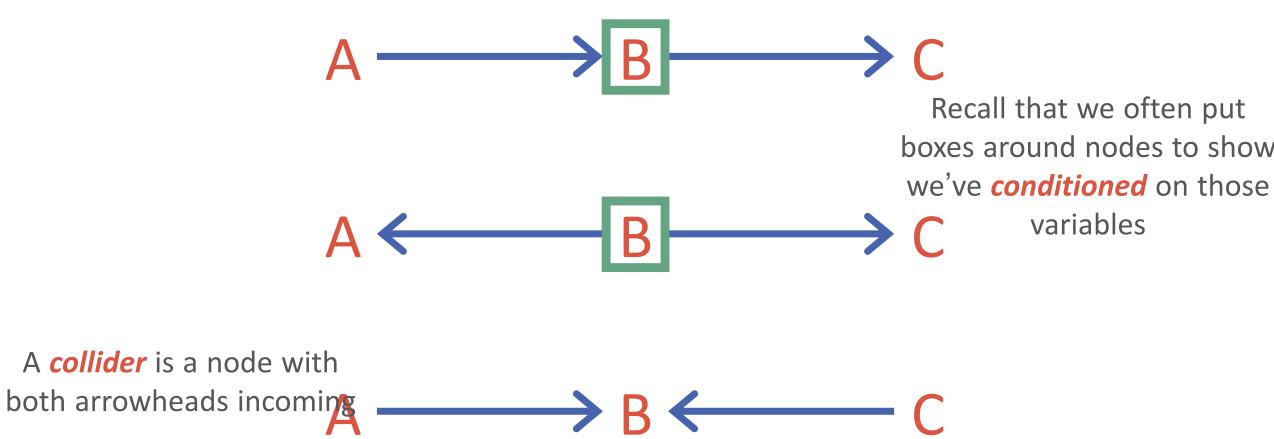
sequence of nodes through time

How do we read graphs?

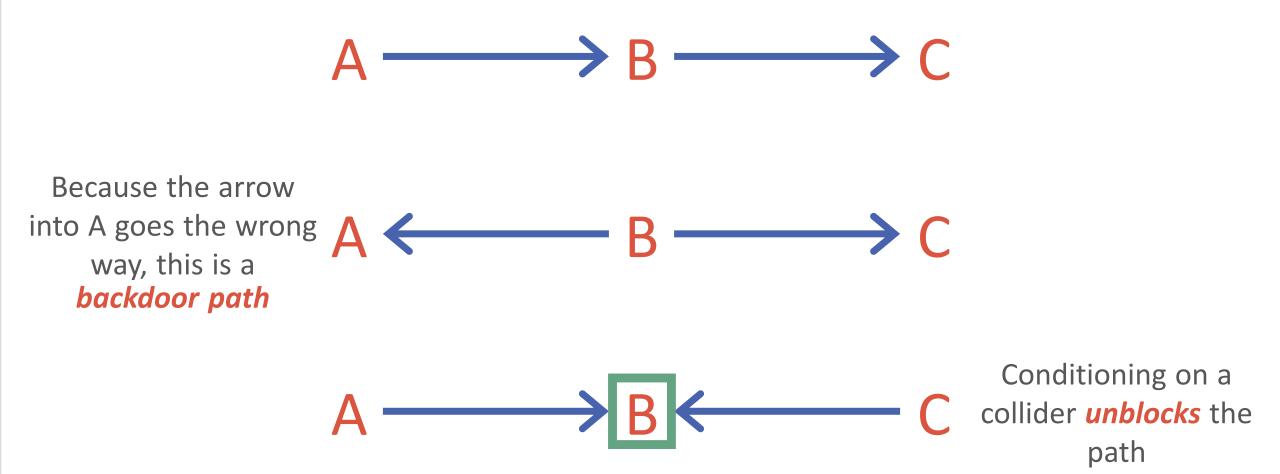
sequences of connected nodes form *paths*



Paths may be *blocked*



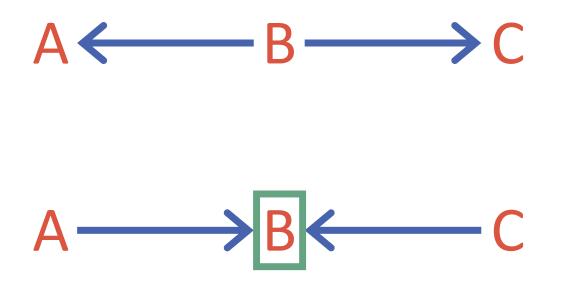




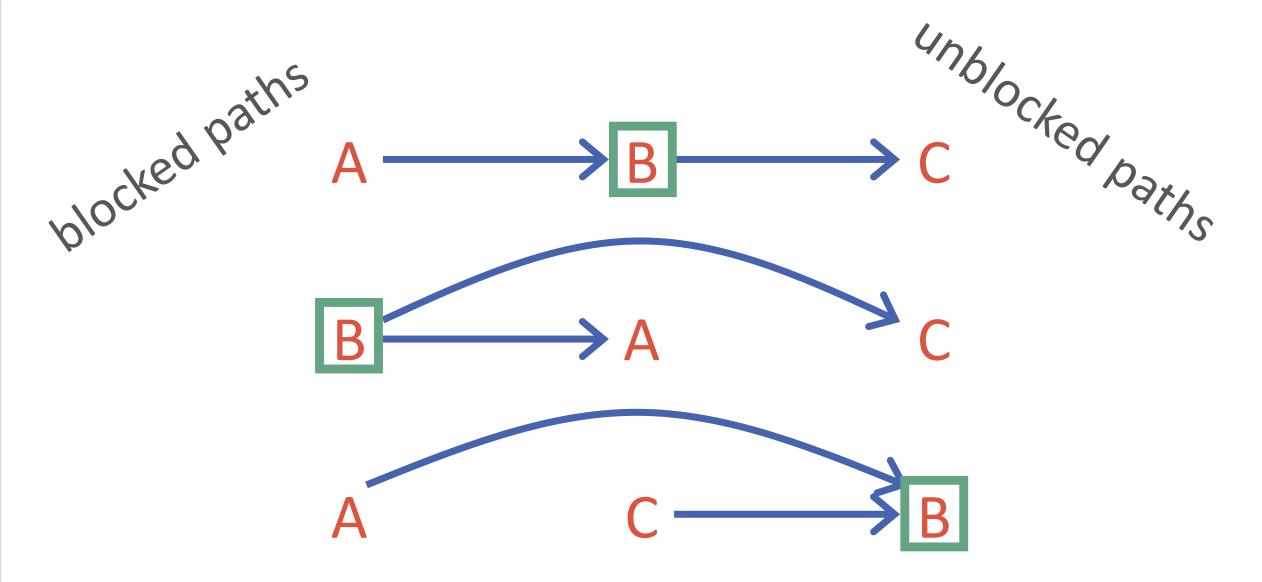
Unblocked paths mean variables are associated

- A and C are *causally* associated
 - We can estimate what will happen to C if we change A
- A and C are non-causally associated
 - Because they're associated in the data, we may think that A causes C
 - But this is just bias!





Graphs are often easier to read in *temporal order*



Directed **DevGs**ic Graphs