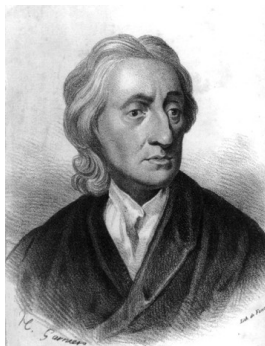


Causality and directed acyclic graphs

Get to know and have some intuition about

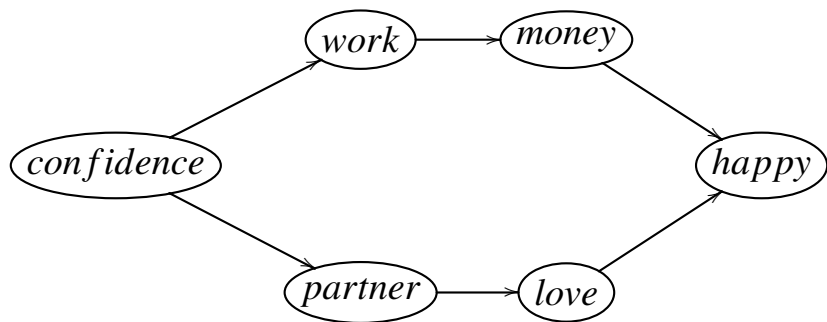
- Causality in philosophy (of science)
- **Conditional independence relations**
- Causal discovery algorithms
- Confounds and back doors



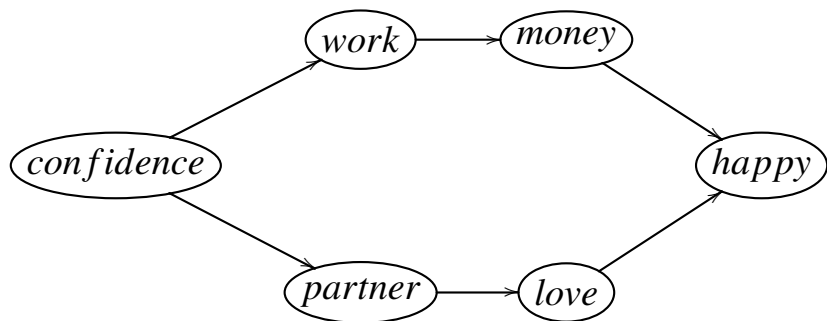
David Hume (1711-1776)

What is a causal graph?

What is a causal graph?



What is a causal graph?

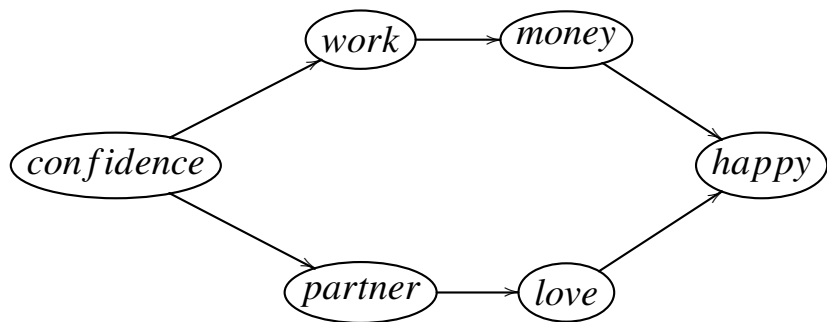


John Stuart Mill

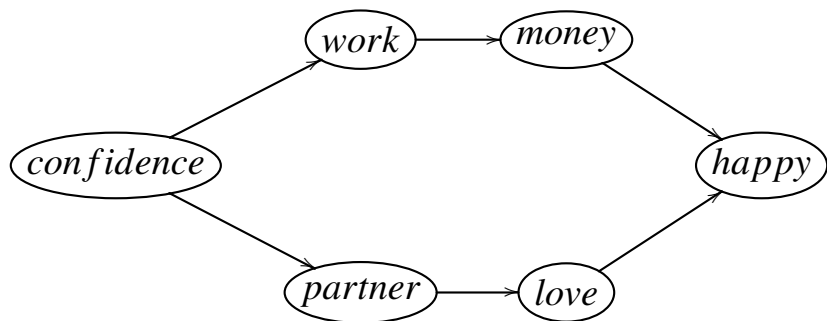
- A (always) co-occurs with B
- A occurs before B
- There is no alternative explanation for the co-occurrence of A and B

What is a causal graph?

What is a causal graph?



What is a causal graph?

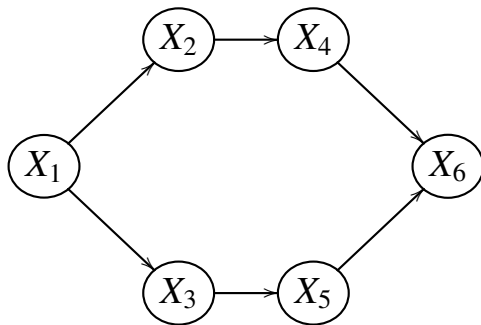


Causality by intervention

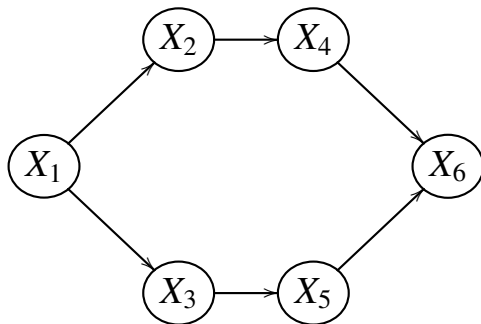
- If I wiggle A then B wiggles too (manipulation)
- There is no alternative explanation for the change in B as a result of the change in A

What is a causal graph?

What is a causal graph?

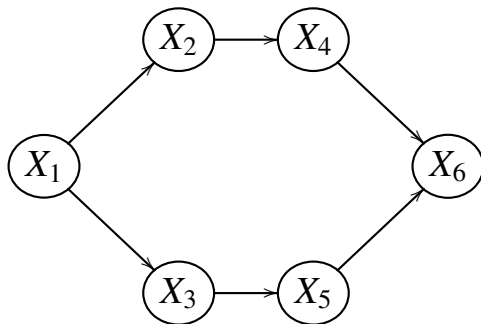


What is a causal graph?



- A directed acyclic graph (DAG)
- A probability distribution over the nodes

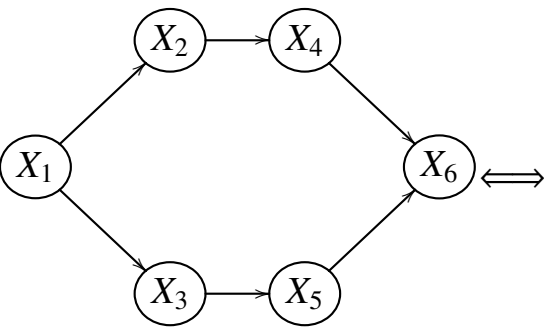
What is a causal graph?



- A directed acyclic graph (DAG)
- A probability distribution over the nodes

This forms the basis to infer causal relations between variables

Graphs and probability



$$\iff P(X_1, X_2, X_3, X_4, X_5, X_6)$$

Graphs and probability

Theorem 1.2.5 (Pearl, 2000, p. 18)

For any three nodes (X, Y, Z) in a DAG G and for all probability functions P , we have

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- (i) if in the graph X and Y are d -separated given Z , then the conditional independence holds in all distributions that are compatible with G ;
and

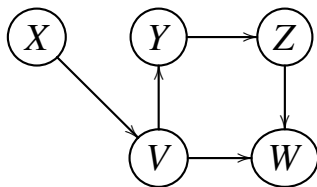
Graphs and probability

Theorem 1.2.5 (Pearl, 2000, p. 18)

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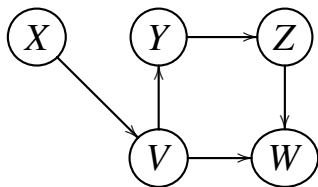
- (i) if in the graph X and Y are d -separated given Z , then the conditional independence holds in all distributions that are compatible with G ; and
- (ii) if X and Y are independent conditional on Z in all distributions compatible with G , then X and Y are d -separated given Z .

d-separation



Lauritzen ($X \perp\!\!\!\perp Z \mid YW$)

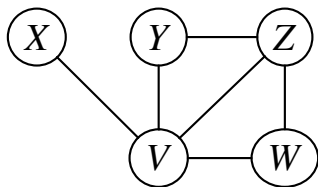
d -separation



Lauritzen ($X \perp\!\!\!\perp Z \mid YW$)

1. Make the *ancestral graph* (G_{an}): the variables of interest and all variables that have a directed path to those variables $\{W, X, Y, Z\}$.

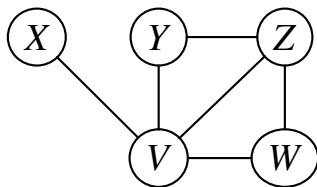
d -separation



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2. Make the *ancestral graph moral* (G_{an}^m): marry all the parents that have a child in common *and* convert all arrows into undirected edges.

d -separation



Lauritzen ($X \perp\!\!\!\perp Z \mid YW$)

1. Make the *ancestral graph* (G_{an}): the variables of interest and all variables that have a directed path to those variables $\{W, X, Y, Z\}$.
2. Make the *ancestral graph moral* (G_{an}^m): marry all the parents that have a child in common *and* convert all arrows into undirected edges.
3. Consider separating all paths in G_{an}^m between X and Z .

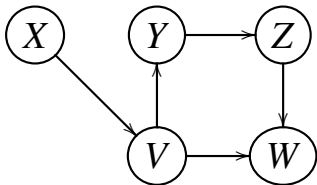
d-separation

$$(X \perp\!\!\!\perp Z \mid YW)$$

d-separation

$$(X \perp\!\!\!\perp Z \mid YW)$$

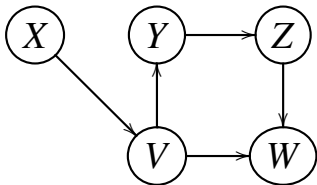
G



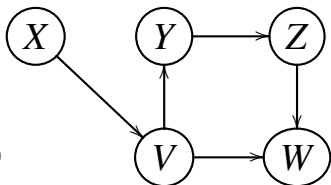
d -separation

$$(X \perp\!\!\!\perp Z \mid YW)$$

G



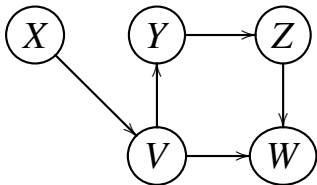
G_{an}



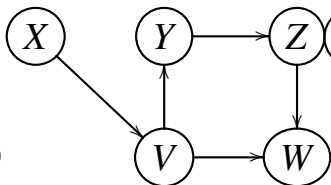
d -separation

$$(X \perp\!\!\!\perp Z \mid YW)$$

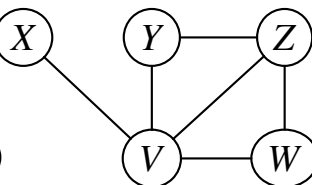
G



G_{an}



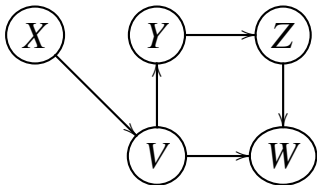
G_{an}^m



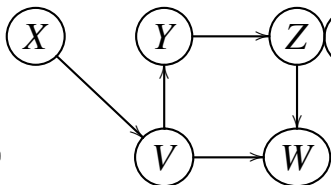
d -separation

$$(X \perp\!\!\!\perp Z \mid YW)$$

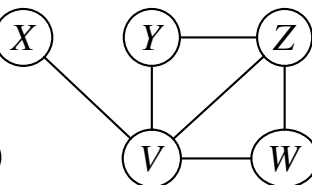
G



G_{an}



G_{an}^m



~~$(X \perp\!\!\!\perp Z \mid YW)$~~

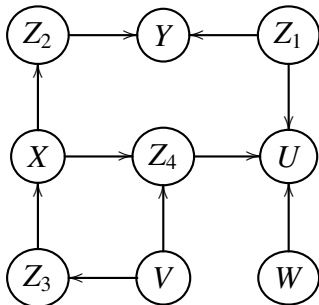
d-separation

$$(X \perp\!\!\!\perp Y \mid Z_2, Z_3)$$

d -separation

$$(X \perp\!\!\!\perp Y \mid Z_2, Z_3)$$

G

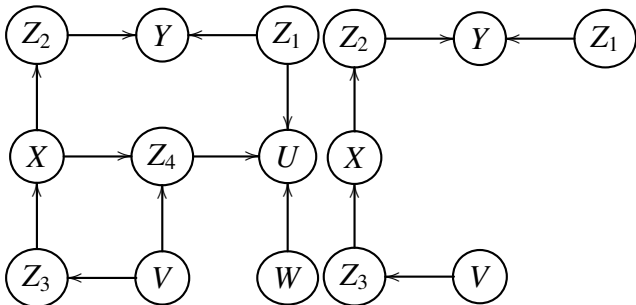


d -separation

$$(X \perp\!\!\!\perp Y \mid Z_2, Z_3)$$

G

G_{an}



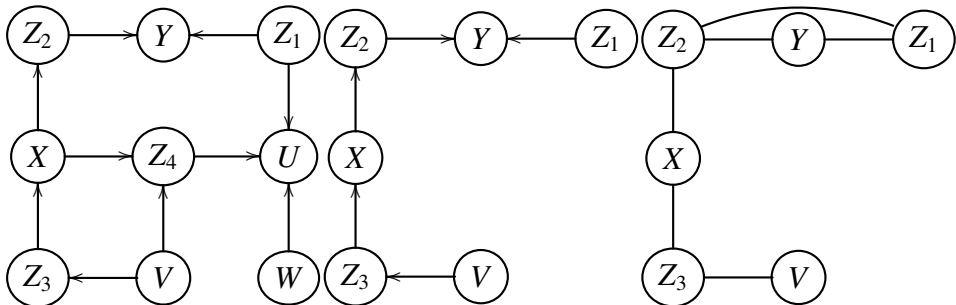
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G

G_{an}

G_{an}^m



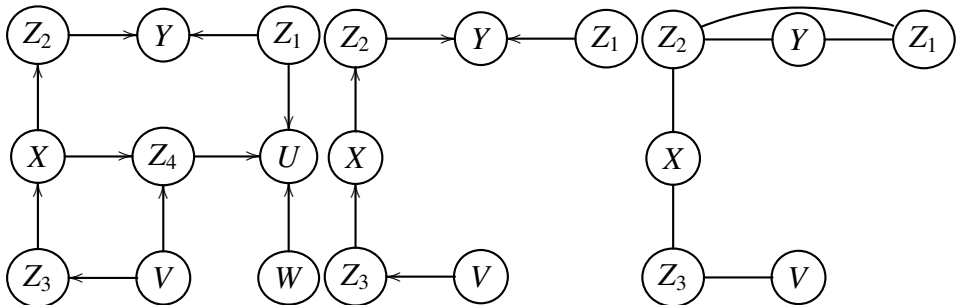
d -separation

$$(X \perp\!\!\!\perp Y \mid Z_2, Z_3)$$

G

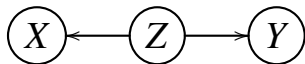
G_{an}

G_{an}^m

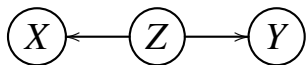


$$(X \perp\!\!\!\perp Y \mid Z_2, Z_3)$$

When are models the same?

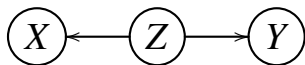


When are models the same?



d -separation

When are models the same?



d -separation

$$(X \perp\!\!\!\perp Y \mid Z)$$

When are models the same?



When are models the same?



$$(X \perp\!\!\!\perp Y \mid Z)$$

When are models the same?



$$(X \perp\!\!\!\perp Y \mid Z)$$

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When are models the same?



$$(X \perp\!\!\!\perp Y \mid Z)$$

$$(X \perp\!\!\!\perp Y \mid Z)$$

$$(X \not\perp\!\!\!\perp Y \mid Z)$$

When are models the same?



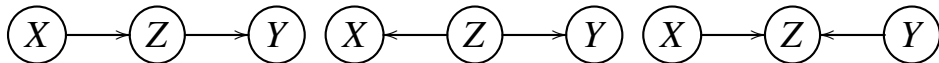
$$(X \perp\!\!\!\perp Y \mid Z)$$

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$$(X \not\perp\!\!\!\perp Y \mid Z)$$

$$(X \perp\!\!\!\perp Y)$$

When are models the same?



$$(X \perp\!\!\!\perp Y \mid Z)$$

$$(X \not\perp\!\!\!\perp Y)$$

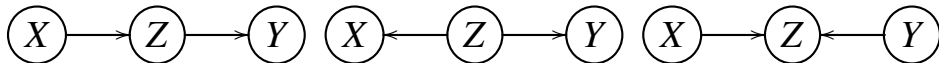
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When are models the same?



$$(X \perp\!\!\!\perp Y \mid Z)$$

$$(X \not\perp\!\!\!\perp Y)$$

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$$(X \not\perp\!\!\!\perp Y)$$

$$(X \not\perp\!\!\!\perp Y \mid Z)$$

$$(X \perp\!\!\!\perp Y)$$

Models are the same when the d -separations are the same!