

The causal structure could be
different from the best posterior
probability structure

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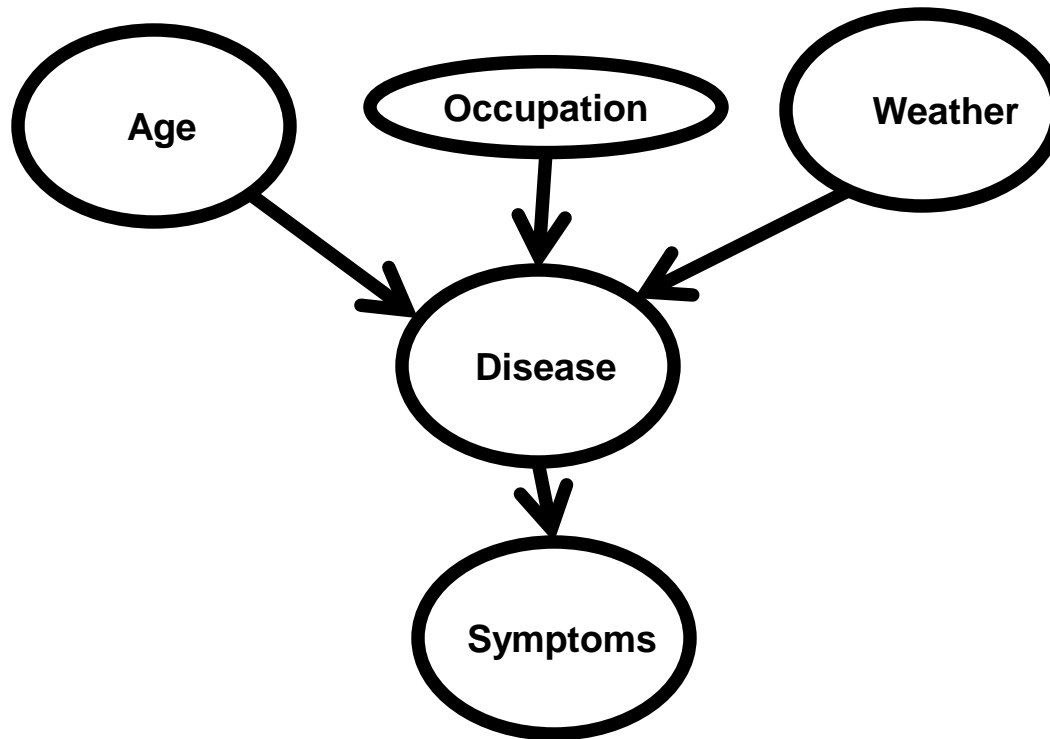


Business school/ finance case

- Keynote address presented at Finance conferences
- Causality used in Investment selection
- Causality can explain Financial distress
- Causality can be used to forecast VaR
- Causality is a tool to explain phenomena

Simplified Bayesian Structure

$P(\text{Age, Occupation, Weather, Disease, Symptoms}) = P(\text{Symptoms} | \text{Disease}) P(\text{Disease} | \text{Age, Occupation, Weather}) P(\text{Age}) P(\text{Occupation}) P(\text{Weather})$
(Buntine, 1996)



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Cooper and Herskovits, 1989

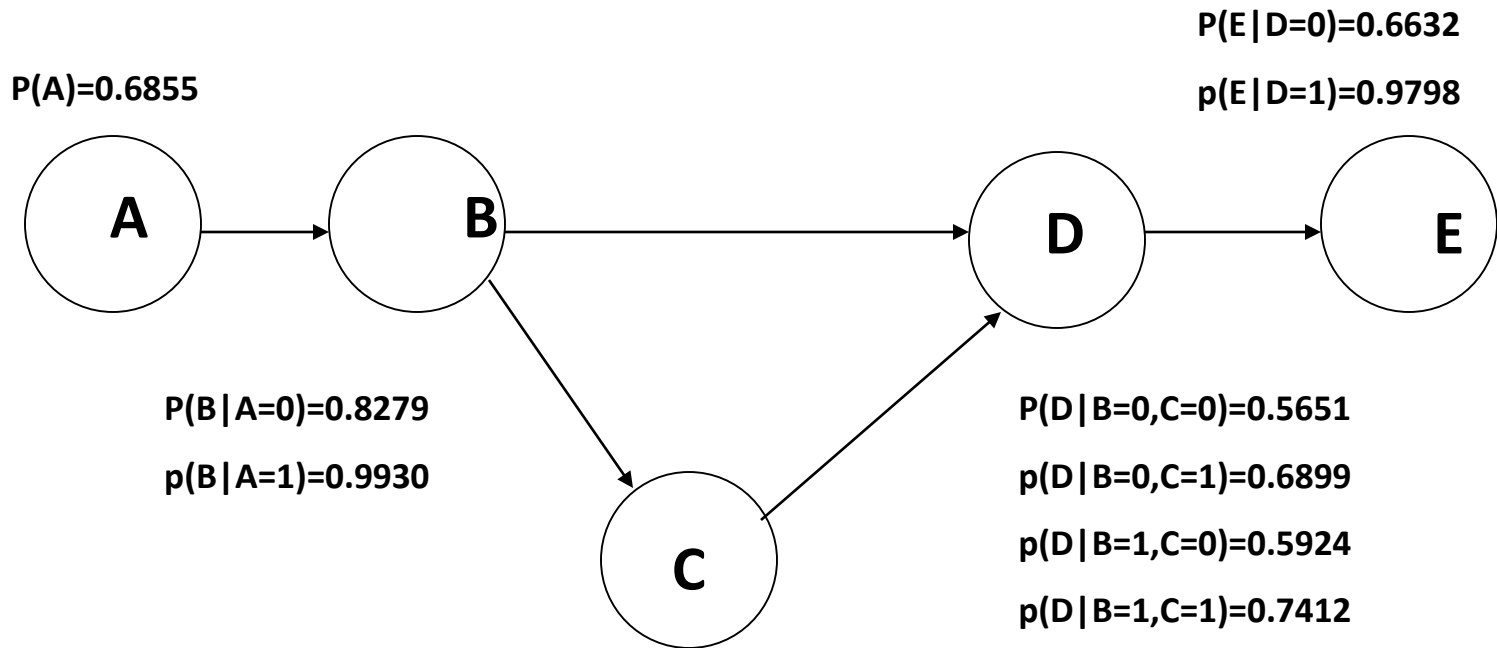
$$p(D | S^h) = \prod_{i=1}^n \prod_{j=1}^q \frac{\Gamma(\alpha_{ij})}{\Gamma(N_{ij} + \alpha_{ij})} \prod_{k=1}^r \frac{\Gamma(N_{ijk} + \alpha_{ijk})}{\Gamma(\alpha_{ijk})}$$

Number of structures

$$3 \frac{n(n-1)}{2}$$

Steps

- Select a model
- Create a frequency distribution
- Do an exhaustive search over structures
- To find the max posterior probability structure
- It must be the same as the structure that created the data.



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Best posterior model is not the causal model

- The unscaled log posterior probability for the correct model is -23755. The models with the best unscaled log posterior probability of -23750 are $A \rightarrow B \rightarrow C \rightarrow D \rightarrow E$, $A \leftarrow B \rightarrow C \rightarrow D \rightarrow E$, $A \leftarrow B \leftarrow C \rightarrow D \rightarrow E$, $A \leftarrow B \leftarrow C \leftarrow D \rightarrow E$ and $A \leftarrow B \leftarrow C \leftarrow D \leftarrow E$. This gives a posterior probability that is e^5 times larger for the wrong model.

New norm

Path between nodes A-B-C

$$\text{Norm } \frac{\rho_{AB}\rho_{BC}}{\rho_{AC}}$$

Given variables A, B and C and assume only linear relationships, then the norm of the path(A-B-C) = 1 if and only if A and C are conditionally independent given B or $I(A, C | B)$

If the norm of the path A-B-C equals one, then

$$\frac{\rho_{AB}\rho_{BC}}{\rho_{AC}} = 1 \quad \text{or} \quad \rho_{AB}\rho_{BC} = \rho_{AC}$$

Only linear relationships are considered,

$$\rho_{AC.B} = \frac{\rho_{AC} - \rho_{AB}\rho_{BC}}{\sqrt{(1 - \rho_{AB}^2)(1 - \rho_{BC}^2)}} = 0$$

therefore A and C are conditionally independent given B or $I(A, C | B)$.

The converse: If A and C are conditionally independent given B, $I(A, C | B)$ then $\rho_{AB}\rho_{BC} = \rho_{AC}$.



Elimination heuristic

- Start with a complete graph containing all the variables.
- Eliminate arcs between variables that are independent (with a correlation coefficient not significantly different from zero).
- Eliminate the arcs where Conditional Independence Statements have been identified (where the norm equals one).
- This gives us the causal tree structure, and I tested that it is also the structure with the highest posterior probability.

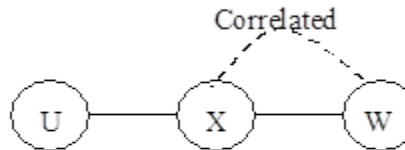
Theorem 1

- If two variables, X and W , are correlated and the path $U-X-W$ is a valid path, then the path with X and W interchanged, $U-W-X$ will have a norm equal to the coefficient of determination (the square of the correlation coefficient) between X and W

If u-x-w valid path then norm of u-w-x = ρ^2



Figure 4.5 Correlated variables on a valid path



Norm(U-X-W) = 1

Norm(U-W-X) = Coeff Det. Between
X and W



If u-x-w valid path then norm of

$$u-w-x = \rho^2$$

- **Proof:** A valid (star decomposable) path U-X-W means that
- ,or, $\rho_{UW} = \rho_{UX} \cdot \rho_{XW}$
- Therefore the norm for the path U-W-X is
- This proves the theorem.
- The following theorem can be used to check if we have found a valid path.

Theorem 2

- X-Y-Z is a valid and only path between X and Z if and only if the product of the norms is equal to the coefficient of determination between the start and end nodes of this path.
- $\text{Norm}(X-Y-Z) \cdot \text{Norm}(Y-X-Z) \cdot \text{Norm}(X-Z-Y) = \rho^2_{XZ}$.

Proof

- Assume a valid and only path X-Y-Z. Then, for a valid and only path there is a tree structure between the variables and the norm will be one (in other words $\rho_{XY}\rho_{XZ} = \rho_{YZ}$).

- Therefore

$$\frac{\rho_{XY}\rho_{YZ}}{\rho_{XZ}} \frac{\rho_{XY}\rho_{YZ}}{\rho_{YZ}} \frac{\rho_{XZ}\rho_{YZ}}{\rho_{XY}} = \rho_{XY}\rho_{XZ}\rho_{YZ} = \rho_{YZ}^2$$

- In the opposite direction: if the product of the norms is equal to the coefficient of determination, then for the path A-E-C this condition can be written as:

$$\frac{\rho_{AE}\rho_{EC}}{\rho_{AC}} \frac{\rho_{AC}\rho_{CE}}{\rho_{AE}} \frac{\rho_{AE}\rho_{AC}}{\rho_{EC}} = \rho_{AC}^2$$

- If the correlation coefficient between A and C is not zero, then
- $\therefore \rho_{EC}\rho_{AE}\rho_{AC} = \rho_{AC}^2$ or A-E-C is a valid and only path.