

Local Semantics and Markov Blanket

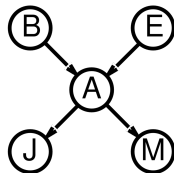
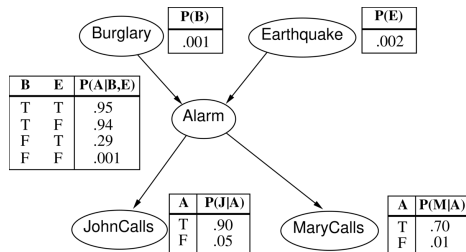


Frederik Mallmann-Trenn
6CCS3AIN

Global semantics

- We can calculate the full joint distribution as the product of the local conditional distributions:

$$P(x_1, \dots, x_n) = \prod_{i=1}^n P(x_i | \text{parents}(X_i))$$



- Compute: $P(j \wedge m \wedge a \wedge \neg b \wedge \neg e)$

Global semantics

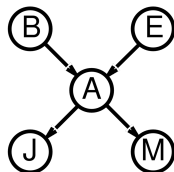
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- $P(j \wedge m \wedge a \wedge \neg b \wedge \neg e)$

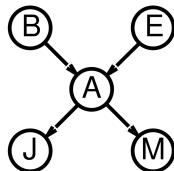
$$\begin{aligned} &= P(j|a)P(m|a)P(a|\neg b, \neg e)P(\neg b)P(\neg e) \\ &= 0.9 \cdot 0.7 \cdot 0.001 \cdot 0.999 \cdot 0.998 \\ &\approx 0.00063 \end{aligned}$$

- Application of the chain rule.



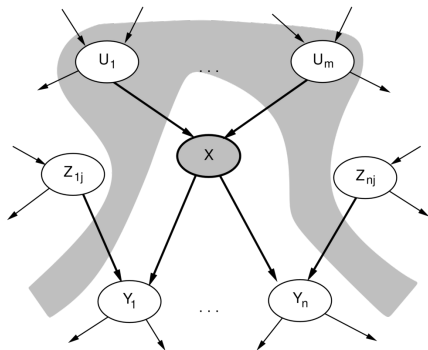
Compactness

- A CPT for Boolean X_i with k Boolean parents has 2^k rows for the combinations of parent values
- Each row requires one number p for $X_i = true$ (the number for $X_i = false$ is just $1 - p$)
- If each variable has no more than k parents, the complete network requires $O(n \cdot 2^k)$ numbers
- Grows linearly with n , vs. $O(2^n)$ for the full joint distribution
- For burglary net, $1 + 1 + 4 + 2 + 2 = 10$ numbers (vs. $2^5 - 1 = 31$)



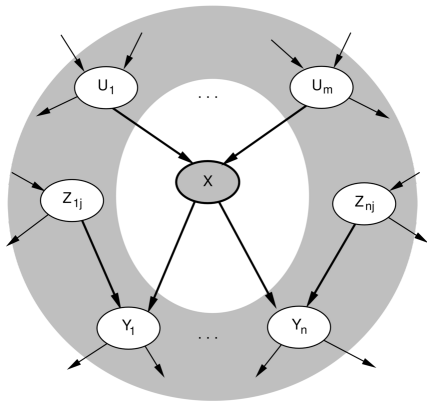
Local semantics

- A node X is conditionally independent of its non-descendants (e.g., the $Z_{i,j}$ s) given its parents (the U_i s shown in the gray area).



Markov blanket

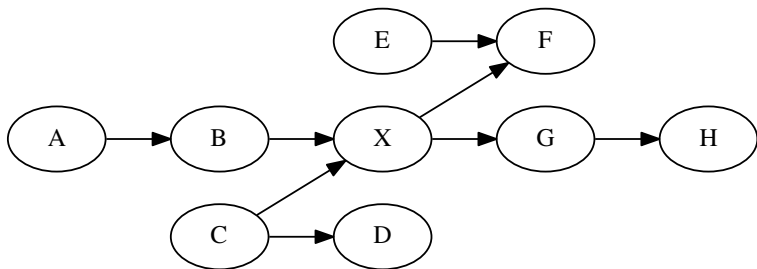
- Each node is conditionally independent of all others given its **Markov blanket**:
parents + children + children's parents



Andrey Markov

Markov blanket

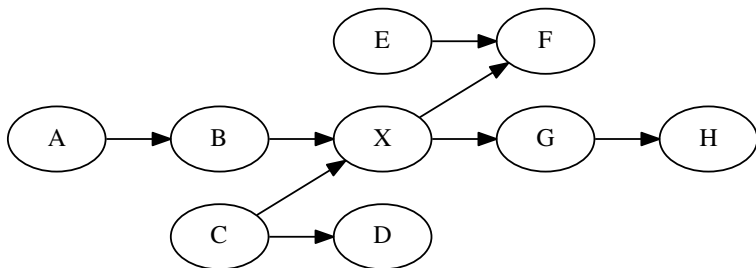
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Markov blanket

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- Markov blanket of X ?
- Answer: B,C,E,F,G

Constructing Bayesian networks

- Build Bayesian networks like any other form of **knowledge representation**.
- First figure out the variables that describe the world.
- Then decide how they are connected.
Conditional independence.
- Then work out the values in the CPTs.



Kathy Laskey

Ways of compressing further

- CPT grows exponentially with number of parents
 - Use distributions that are defined compactly
- **Deterministic** nodes are the simplest case.
- $X = f(\text{Parents}(X))$ for some function f
 - Boolean functions:

$$\text{NorthAmerican} \Leftrightarrow \text{Canadian} \vee \text{US} \vee \text{Mexican}$$

- Numerical relationships among continuous variables

$$\frac{\partial \text{Level}}{\partial t} = \text{inflow} + \text{precipitation} - \text{outflow} - \text{evaporation}$$