Lecture 22: 21 April, 2022

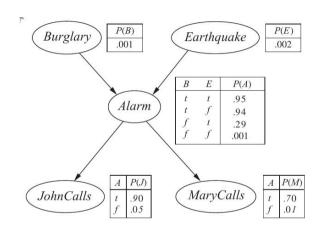
Madhavan Mukund

https://www.cmi.ac.in/~madhavan

Data Mining and Machine Learning January–May 2022

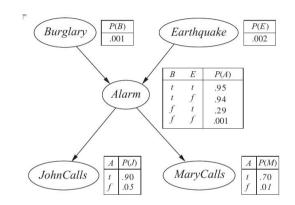
Probabilistic graphical models

- Underlying DAG, no cyclic dependencies
- Each node has a local (conditional) probability table

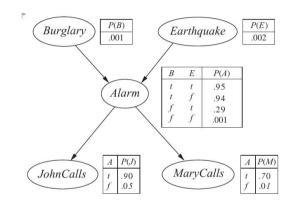


 \blacksquare $x \perp y - x$ and y are independent

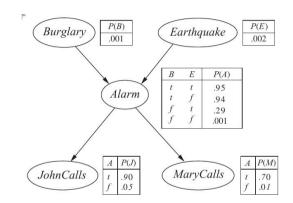
$$P(x \wedge y) = P(x) \cdot P(y)$$



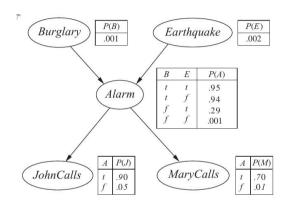
- $\blacksquare x \perp y x$ and y are independent
 - $P(x \land y) = P(x) \cdot P(y)$
- $\blacksquare x \perp y \mid z$
 - x and y are independent given z



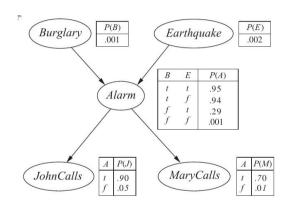
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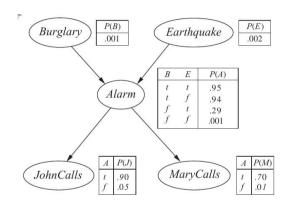
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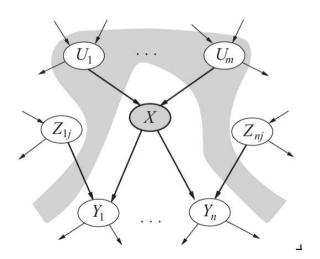
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 - Yes by semantics of network, local independence



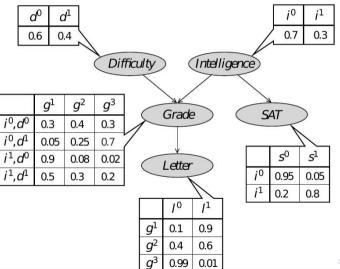
Probabilistic graphical models

■ Fundamental assumption

A node is conditionally independent of non-descendants, given its parents

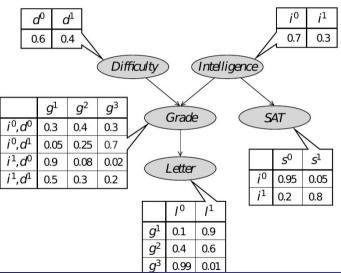


■ *SAT* ⊥ *Grade* | *Difficulty* ?

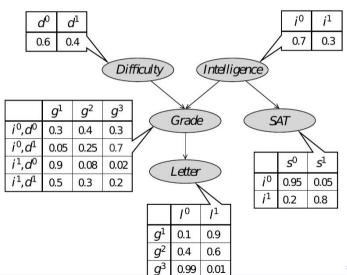


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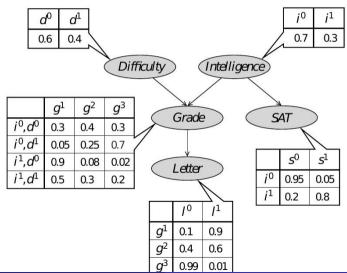
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- Can we calculate conditional independence from the graph?



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- Can we calculate conditional independence from the graph?
- In general, check if X ⊥ Y | Z for sets of variables X, Y, Z

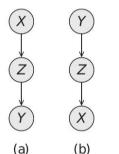


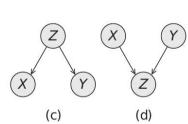
How does dependence "flow" through a network?

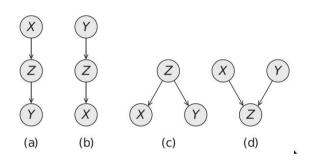
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- Examine trails between nodes
 - Paths in the underlying undirected graph

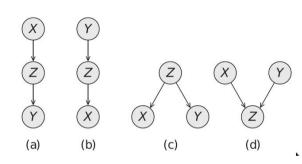
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- Basic trails (undirected) paths of length 2
 - Four basic trails



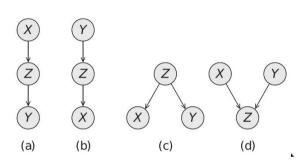




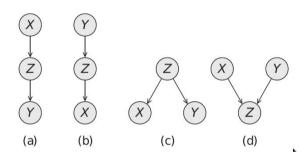
■ (a), (b) and (c): Z blocks flow between X and Y, by semantics of Bayesian networks



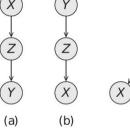
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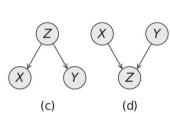


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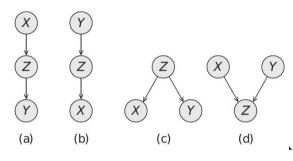


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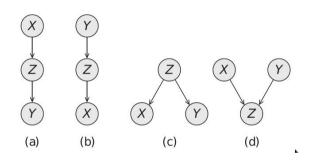




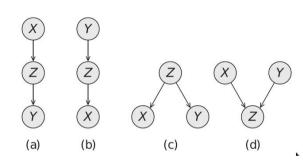
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 - Simplest form of V-structure



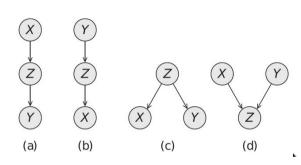
• Check if $X \perp Y \mid Z$



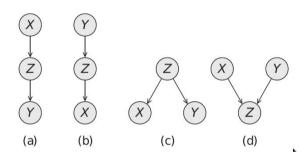
- Check if $X \perp Y \mid Z$
- Dependence should be blocked on every trail from X to Y



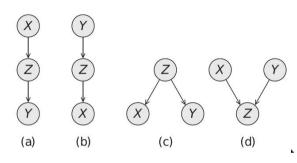
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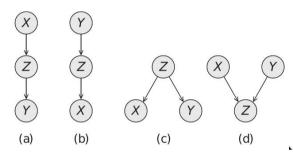
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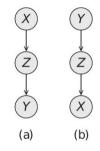
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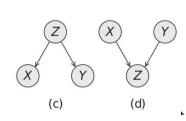


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 - In general, V-structure includes descendants of the bottom node



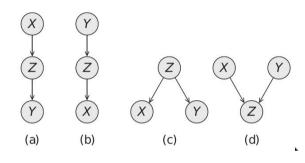
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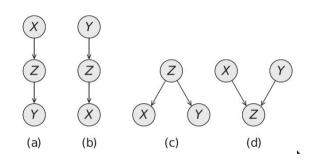
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- Variation of breadth first search (BFS) to check if *y* is reachable from *x* through some trail

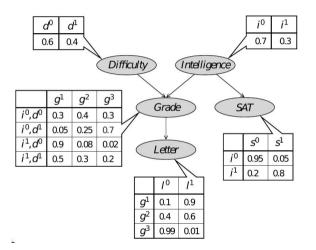
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- x and y are D-separated given z if all trails are blocked
- Variation of breadth first search (BFS) to check if y is reachable from x through some trail
- Extends to sets each $x \in X$ is D-separated from each $y \in Y$

Conditional independence, example

- Is SAT independent of Difficulty given Intelligence?
 - Yes, Difficulty Grade Intelligence - SAT trail is blocked at Grade (V-structure) and Intelligence



Conditional independence, example

- Is SAT independent of Difficulty given Intelligence?
 - Yes, Difficulty Grade Intelligence
 SAT trail is blocked at Grade
 (V-structure) and Intelligence
- Is SAT independent of Difficulty given Grade?
 - No, Difficulty Grade Intelligence
 SAT trail is open because Grade
 is known (V-structure) and
 Intelligence is not

