

# Lecture 22: 4 April, 2023

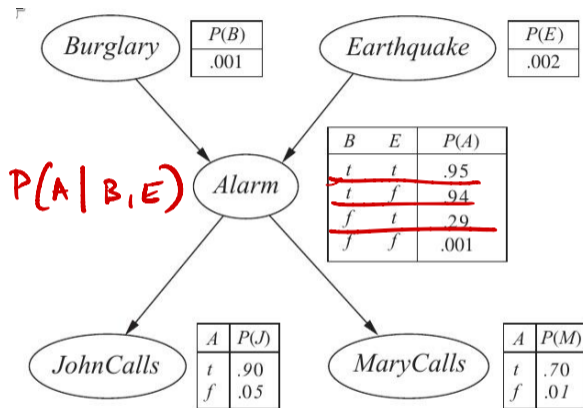
Madhavan Mukund

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

Data Mining and Machine Learning  
January–April 2023

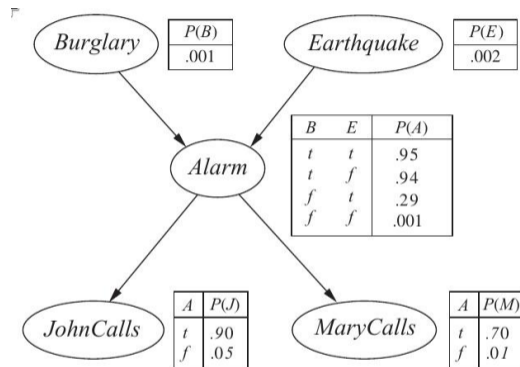
# Probabilistic graphical models

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



# Conditional independence

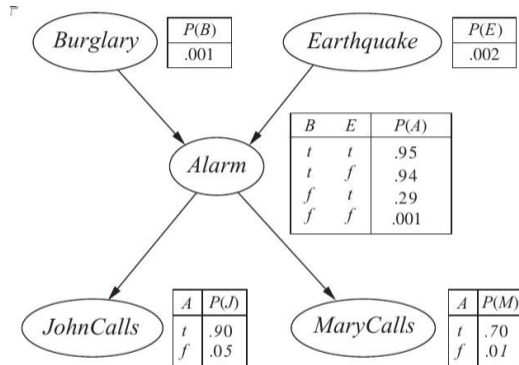
- $x \perp y$  —  $x$  and  $y$  are independent
  - $P(x \wedge y) = P(x) \cdot P(y)$



# Conditional independence

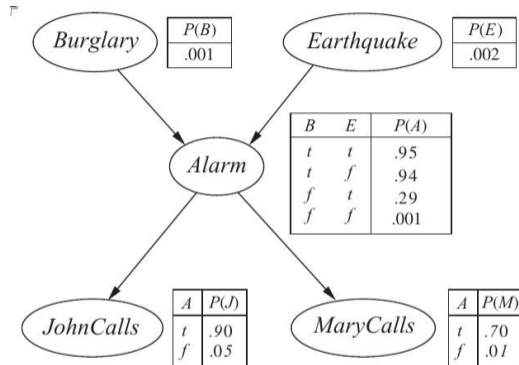
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- $x \perp y \mid z$ 
  - $x$  and  $y$  are independent given  $z$
  - $P(x \wedge y \mid z) = P(x \mid z) \cdot P(y \mid z)$

$$P(x \mid y, z) = P(x \mid z)$$
$$P(y \mid x, z) = P(y \mid z)$$



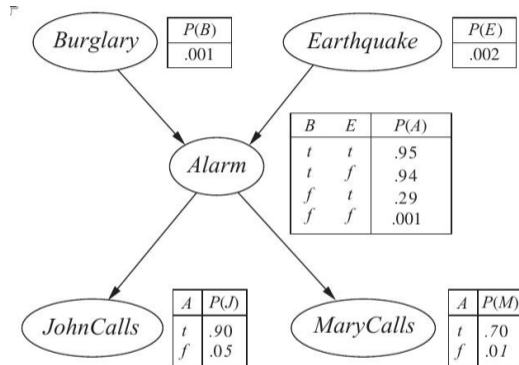
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- Is *JohnCalls* independent of *MaryCalls* ( $j \perp m$ )?



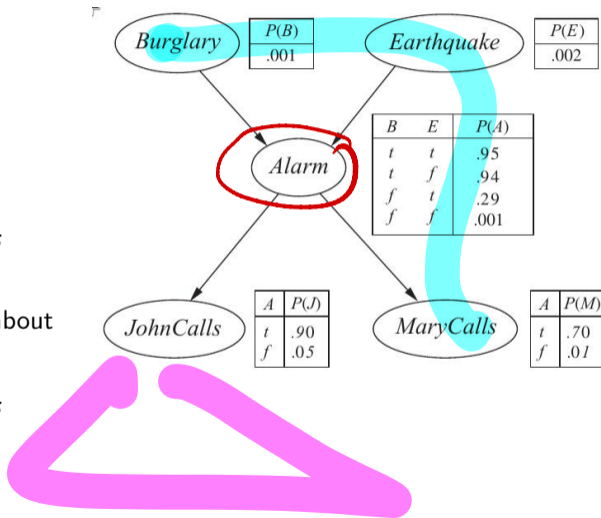
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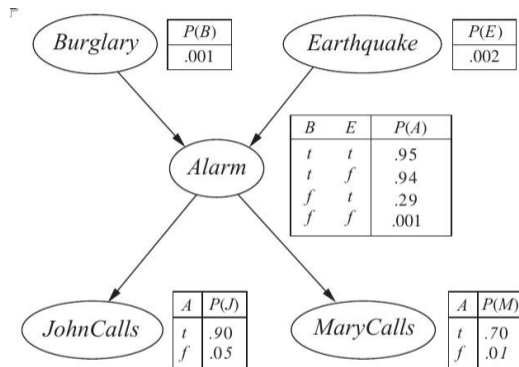
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  - Yes — by semantics of network, local independence

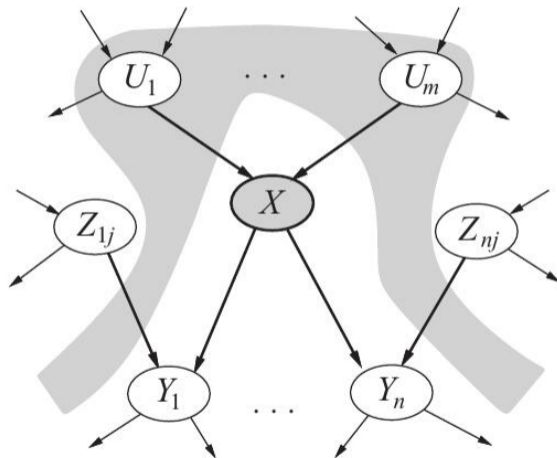




# Probabilistic graphical models

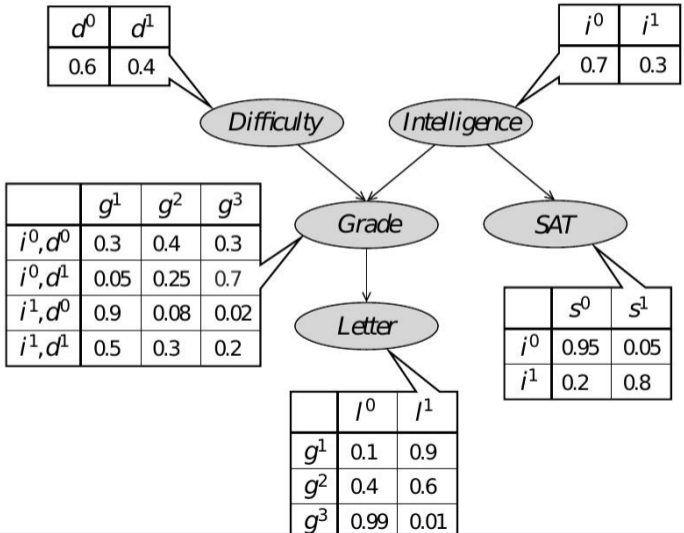
## ■ Fundamental assumption

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



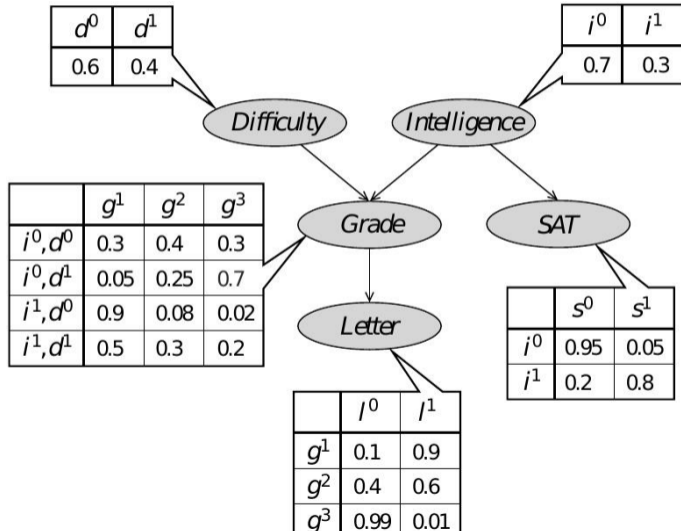
# Student example

■  $SAT \perp Grade \mid Difficulty ?$



# Student example

- $SAT \perp Grade \mid Difficulty$  ?
  - No

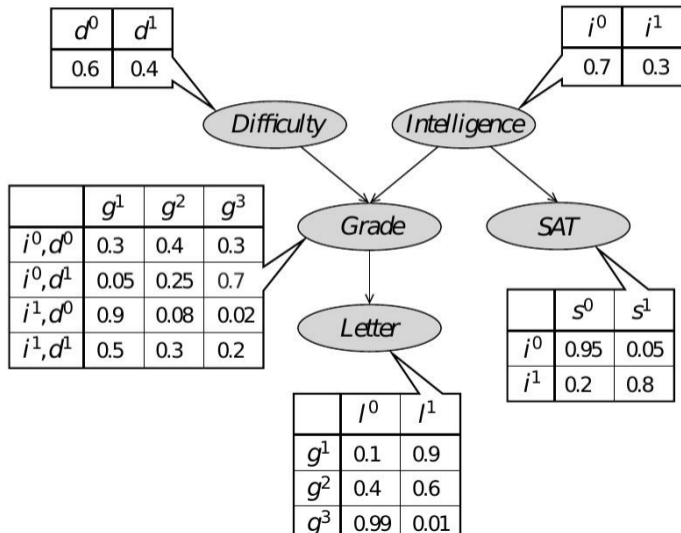


# Student example

- $SAT \perp Grade \mid Difficulty$  ?

- No

- Can we calculate conditional independence from the graph?



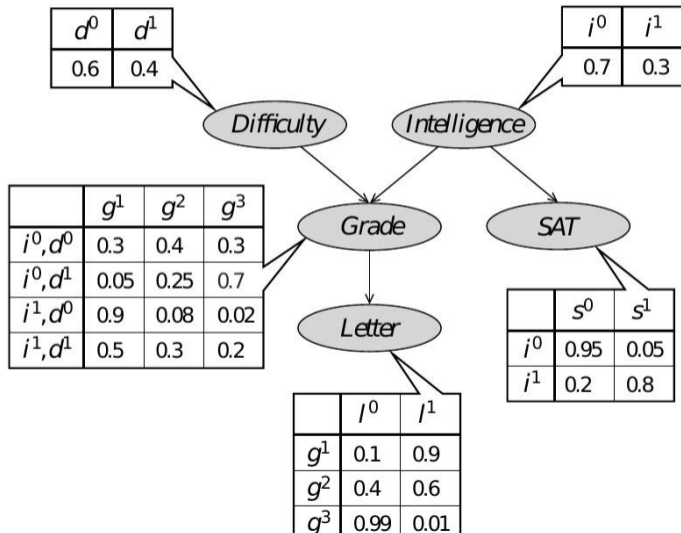
# Student example

■  $SAT \perp Grade \mid Difficulty$  ?

■ No

■ Can we calculate conditional independence from the graph?

■ In general, check if  $X \perp Y \mid Z$  for sets of variables  $X, Y, Z$



# Conditional independence

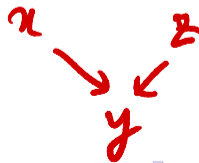
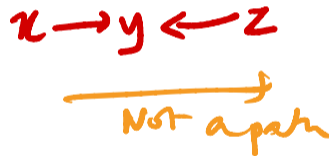
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# Conditional independence

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- For neighbouring nodes, dependence flows both ways
  - If  $x \rightarrow y$ , knowing  $x$  tells us about  $y$  and vice versa

# Conditional independence

- How does dependence “flow” through a network?
- For neighbouring nodes, dependence flows both ways
  - If  $x \rightarrow y$ , knowing  $x$  tells us about  $y$  and vice versa
- Examine **trails** between nodes *Ignore direction.*
  - Paths in the underlying undirected graph





# Conditional independence

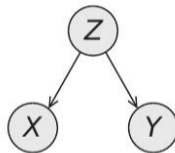
- How does dependence “flow” through a network?
- For neighbouring nodes, dependence flows both ways
  - If  $x \rightarrow y$ , knowing  $x$  tells us about  $y$  and vice versa
- Examine **trails** between nodes
  - Paths in the underlying undirected graph
- **Basic trails** — (undirected) paths of length 2
  - Four basic trails



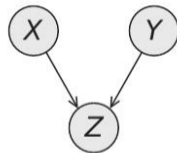
(a)



(b)



(c)



(d)

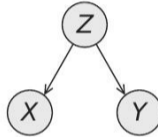
# Basic trails



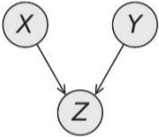
(a)



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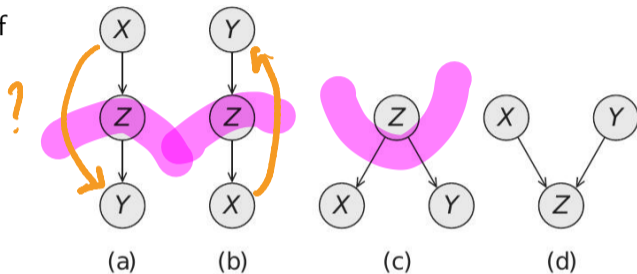
(c)



(d)

# Basic trails

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



# Basic trails

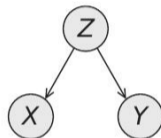
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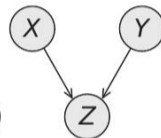
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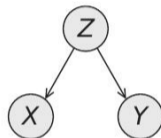
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  - $Z$ : Car does not start
  - $X$ : Low Battery,  $Y$ : No Fuel



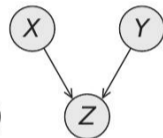
(a)



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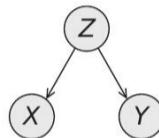
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 $X$ : Overnight rain,  $Y$ : Sprinkler ran



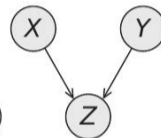
(a)



(b)



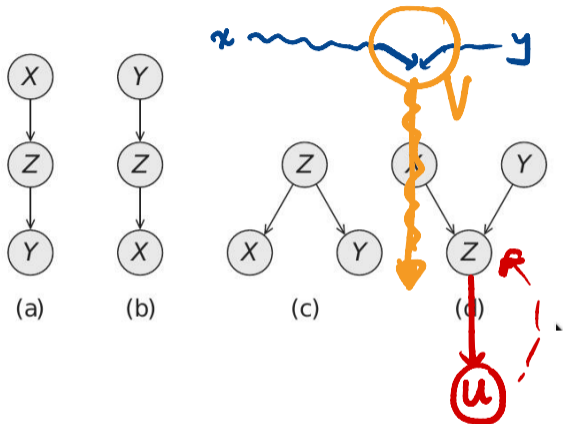
(c)



(d)

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  - Simplest form of  $V$ -structure



$$X \perp Y \mid U ?$$
$$\approx X \perp Y \mid Z$$

# D-Separation

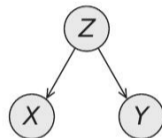
- Check if  $X \perp Y \mid Z$



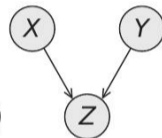
(a)



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(d)



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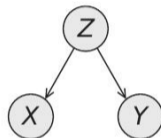
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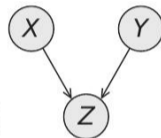
(a)



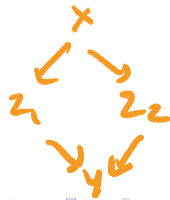
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(d)



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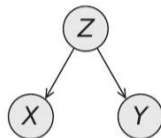
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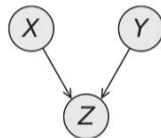
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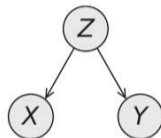
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  - For (a), (b), (c), need  $Z$  present



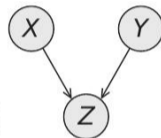
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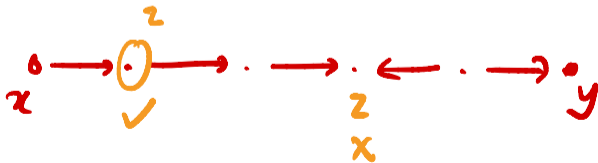
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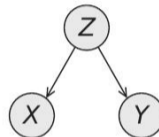
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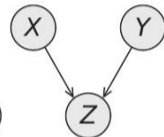
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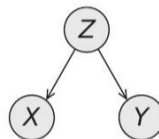
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  - In general, V-structure includes descendants of the bottom node



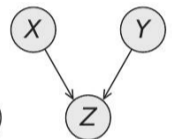
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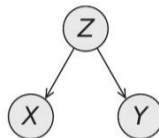
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  - In general, V-structure includes descendants of the bottom node
- $x$  and  $y$  are **D-separated** given  $z$  if all trails are blocked



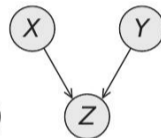
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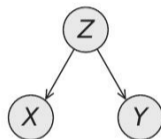
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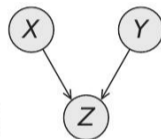
(a)



(b)



(c)



(d)

- $x$  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

# Breadth First Search

Reachability in a graph

Is  $y$  reachable from  $x$ ?

- Mark all neighbours of  $x$  as reachable

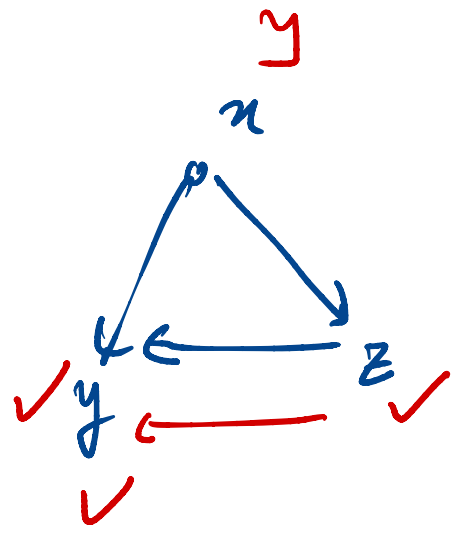
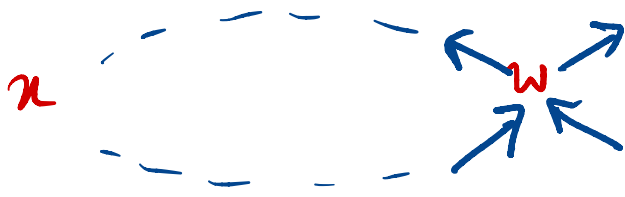
- Mark nbrs of nbrs

⋮

No new marks

Is  $y$  marked?





# D-Separation

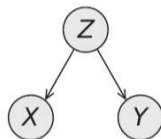
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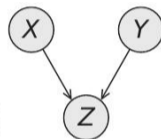
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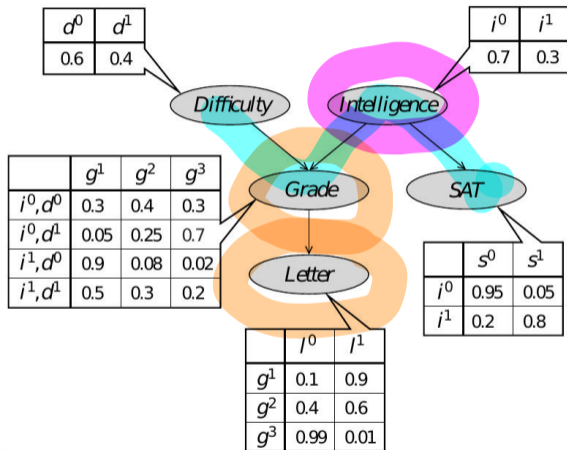


(d)

- $x$  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 **Letter**?
  - No, **Difficulty** – **Grade** – **Intelligence** – **SAT** trail is open
  - **Letter** is known, hence something about **Grade** is known (V-structure)
  - **Intelligence** is not known

