

Quiz 2: due Fri 11:59pm  
 HW 1 posted, due Fri 9/27 11:59pm. Should cover everything  
 COMPSCI 688: Probabilistic Graphical Models by Wed.  
 Lecture 4: Directed Graphical Models: D-Separation, Queries

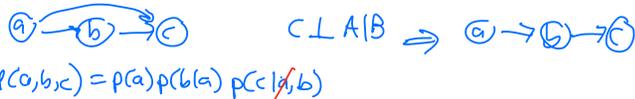
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Partially based on materials by Benjamin M. Marlin (marlin@cs.umass.edu) and Justin Domke (domke@cs.umass.edu)

Review

Review



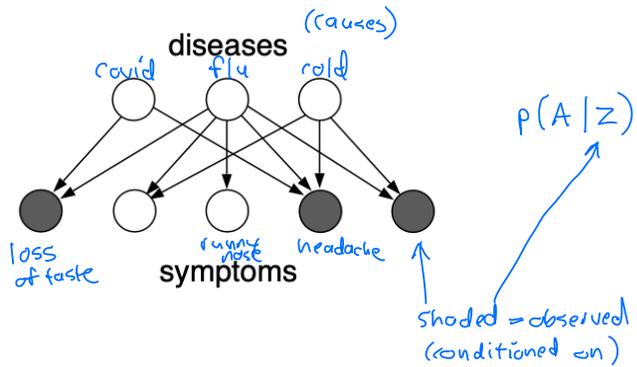
- ▶ Bayes net:  $p(\mathbf{x}) = \prod_{i=1}^N p(x_i | \mathbf{x}_{pa(i)})$
- ▶ Factorization  $\iff$  conditional independence (one statement per node)

$$p(\mathbf{x}) = \prod_{i=1}^N p(x_i | \mathbf{x}_{pa(i)}) \iff X_i \perp \mathbf{X}_{nd(i)} | \mathbf{X}_{pa(i)} \text{ for all } i$$

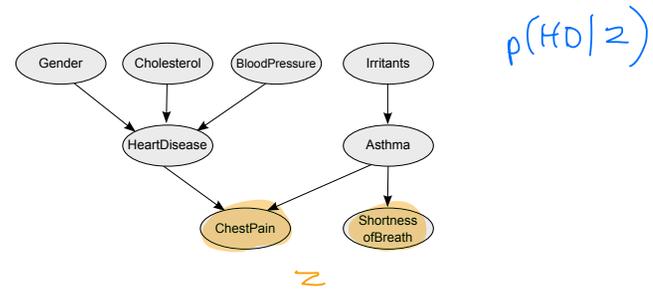
- ▶ We would like to chain together conditional independence properties using the graph structure to derive new ones  $\rightarrow$  **D-separation**

Examples

### Example 1: Medical Diagnosis "Expert Systems"

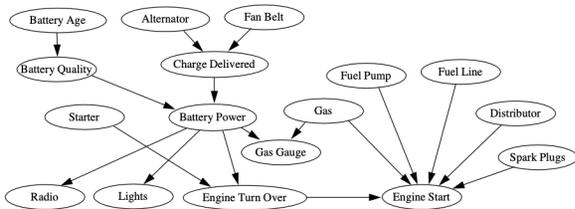


### Example 2: Medical Diagnosis

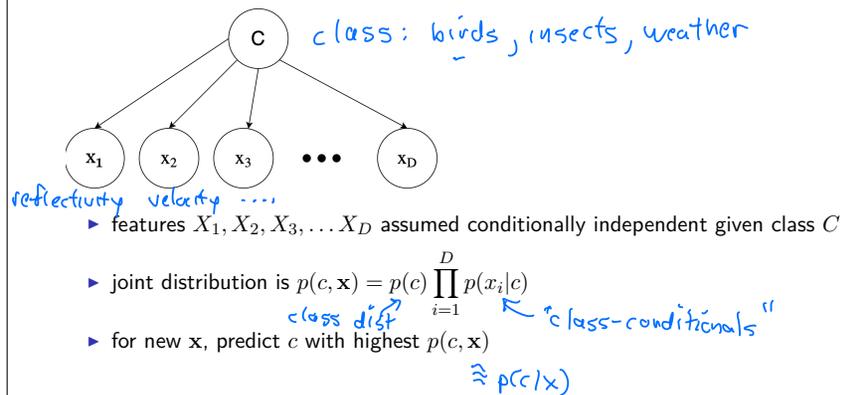


### Example 3: Equipment Diagnostics

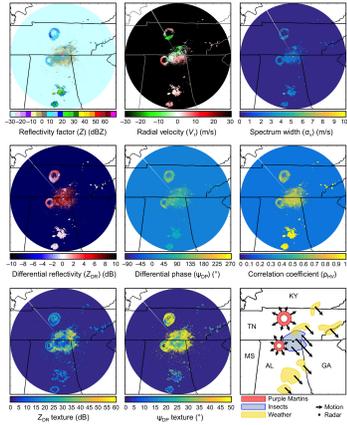
Suppose engine does not start but radio plays? What do you believe about gas in tank?



### Example 4: Naive Bayes

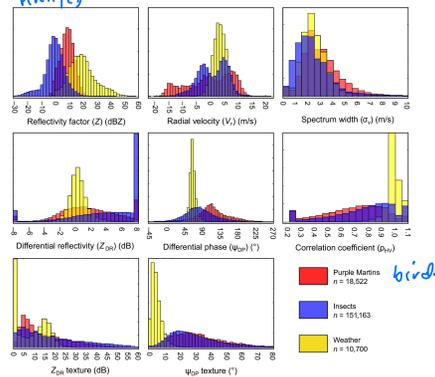


### Radar Example



Dual-polarization radar products for biological applications, Ecosphere, Volume: 7, Issue: 11, First published: 07 November 2016, DOI: (10.1002/ecs2.1539)

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### Mechanistic and Statistical Modeling

- ▶ COVID model
  - ▶ 8-schools model
- } later

### Warning: Causality

#### Bayes nets are not causal

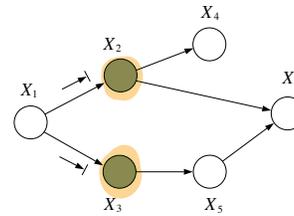
- ▶ Many of our examples are motivated by a causal model, but Bayes net arrows could just as easily point from “effect” to “cause” (or have no causal semantics at all)
- ▶ Can be given causal semantics (beyond our scope)



### D-Separation

### Independence Properties

So far, we know  $X_i \perp X_{nd(i)} | X_{pa(i)}$  for all  $i$



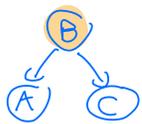
- 1.  $X_1 \perp \emptyset | \emptyset$
- 2.  $X_2 \perp X_3, X_5 | X_1$
- 3. ...
- 6.  $X_6 \perp X_1, X_3, X_4 | X_2, X_5$

However, this also implies other conditional independence properties. E.g., it's true that  $X_1 \perp X_6 | X_2, X_3$  in this network. How can we determine this?

The core principles can be understood by examining three-node networks, then "chaining" ideas together...

### Three-Node Bayes Nets: Common Parent, Chains

Networks  $A \leftarrow B \rightarrow C$ ,  $A \rightarrow B \rightarrow C$  and  $C \rightarrow B \rightarrow A$   
"common cause" "causal chain"



$A \not\perp C$   
 $A \perp C | B$



$A \perp C$   
 $A \perp C | B$

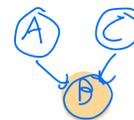


$A \perp C$   
 $A \perp C | B$

$A \not\perp C$  but  $A \perp C | B$ . Observing  $B$  blocks dependence of  $A$  and  $C$

### Three-Node Bayes Nets: V-Structure

Network  $A \rightarrow B \leftarrow C$



causes  
"common effect"

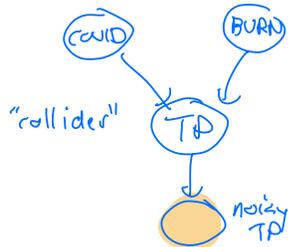
"V-structure"

$A \perp C$   
 $A \not\perp C | B$

$A \perp C$  but  $A \not\perp C | B$ . Observing  $B$  induces dependence of  $A$  and  $C$

### Explaining Away

- ▶ "Explaining away" via V-structures is a distinguishing property of Bayes nets:
- ▶ **Example:** You have tongue pain and loss of sensation. Do you have COVID or did you burn your tongue?



$$P(\text{Covid}) = .001$$

$$P(\text{Covid} | \text{TP}) = .002$$

$$P(\text{Covid} | \text{TP}, \text{Burn}) = .0011$$

Covid  $\perp$  Burn | TP

In words: if there are two possible causes for the observed evidence, knowing about one of the causes provides information about the other

### D-Separation

**Directed separation** or **D-separation** is a definition of separation in a directed graph that corresponds exactly to conditional independence in Bayes nets

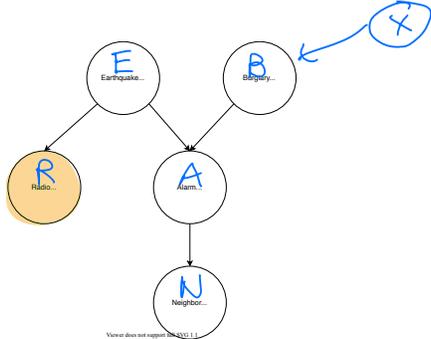
A three-node path is blocked iff has one of the following types:

- 1)  $A \rightarrow B \rightarrow C$  or  $C \rightarrow B \rightarrow A$  and  $B$  is observed
- 2)  $A \leftarrow B \rightarrow C$  and  $B$  is observed
- 3)  $A \rightarrow B \leftarrow C$  and neither  $B$  nor any descendent of  $B$  is observed

Let  $X$ ,  $Y$ , and  $Z$  be three sets of nodes.  $X$  and  $Y$  are d-separated given observed nodes  $Z$  iff every path from  $X$  to  $Y$  is blocked, where a path is blocked if any three-node sequence in the path is blocked.

~~clearly~~  $X$  and  $Y$  are d-separated given  $Z \iff X \perp Y | Z$

### Example: D-separation in the Alarm Model



$$E \perp X | A, B? \text{ YES}$$

$$E \perp X | A, R? \text{ NO}$$

$$E \perp B | A, R? \text{ NO}$$

$$E \perp B | R, N? \text{ NO}$$

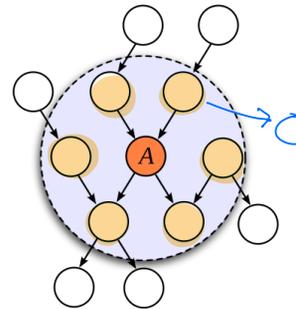
$$E \perp B | R? \text{ YES}$$

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### Markov Blanket

Ex:  $P(A | \underbrace{B, C, D, E, F, \dots, Z}_{\text{all nodes}}) = P(A | \underbrace{C, F, G}_{\text{MB}})$

A Markov blanket of  $A$  is a set of nodes that d-separates  $A$  from the remaining nodes.



In a Bayes net, a Markov blanket of  $A$  consists of:

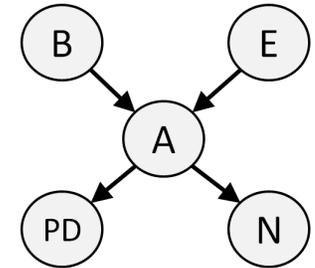
- ▶ parents of  $A$
- ▶ children of  $A$
- ▶ parents of children of  $A$

## Queries

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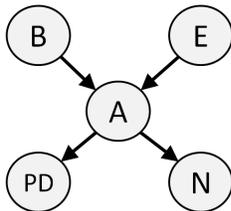
## The Alarm Network (II)

- ▶ You live in the suburbs of LA. Your home alarm may go off because of a break-in or earthquake. If your alarm goes off you might get a call from the police or your neighbor.
- ▶ **Random Variables:** Break-in (B), Earthquake (E), Alarm (A), Police Department calls (PD), Neighbor calls (N).



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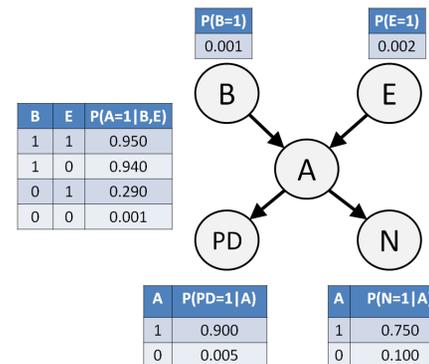
## The Alarm Network: Factorization



- ▶ **Factorization:**  $P(B, E, A, PD, N) = P(B)P(E)P(A|B, E)P(PD|A)P(N|A)$

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## The Alarm Network: Parameters

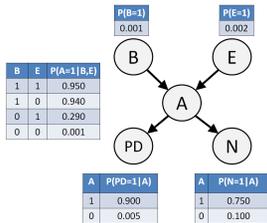


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## The Alarm Network: Joint Query

- ▶ **Question:** What is the probability that there is a break-in, but no earthquake, the alarm goes off, the police call, but your neighbor does not call?

$$\begin{aligned} P(B=1, E=0, A=1, PD=1, N=0) \\ &= P(B=1)P(E=0)P(A=1|B=1, E=0)P(PD=1|A=1)P(N=0|A=1) \\ &= 0.001 \cdot (1 - 0.002) \cdot 0.94 \cdot 0.9 \cdot (1 - 0.75) \end{aligned}$$



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## The Alarm Network: Marginal Query

- ▶ **Question:** What is the probability that there was a break-in, but no earthquake, the police call, but your neighbor does not call?

$$\begin{aligned} P(B=1, E=0, PD=1, N=0) \\ &= \sum_{a=0}^1 P(B=1, E=0, A=a, PD=1, N=0) \\ &= P(B=1)P(E=0)P(A=1|B=1, E=0)P(PD=1|A=1)P(N=0|A=1) \\ &\quad + P(B=1)P(E=0)P(A=0|B=1, E=0)P(PD=1|A=0)P(N=0|A=0) \\ &= 0.001 \cdot (1 - 0.002) \cdot 0.94 \cdot 0.9 \cdot (1 - 0.75) \\ &\quad + 0.001 \cdot (1 - 0.002) \cdot (1 - 0.94) \cdot 0.005 \cdot (1 - 0.1) \end{aligned}$$

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## The Alarm Network: Conditional Query

- ▶ **Question:** What is the probability that the alarm went off given that there was a break-in, but no earthquake, the police call, but your neighbor does not call?

$$\begin{aligned} P(A=1|B=1, E=0, PD=1, N=0) \\ &= \frac{P(B=1, E=0, A=1, PD=1, N=0)}{\sum_{a=0}^1 P(B=1, E=0, A=a, PD=1, N=0)} \\ &= \frac{P(B=1)P(E=0)P(A=1|B=1, E=0)P(PD=1|A=1)P(N=0|A=1)}{\sum_{a=0}^1 P(B=1)P(E=0)P(A=a|B=1, E=0)P(PD=1|A=a)P(N=0|A=a)} \end{aligned}$$

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## The Alarm Network: More Queries

- ▶ What is the probability that there is a break-in given that there is an earthquake?
- What is the probability that your neighbor calls given that the alarm goes off and there is an earthquake?
- What is the probability that the police call given that the alarm goes off and your neighbor calls?
- What is the probability of a break-in given that the alarm goes off and the police call?
- What is the probability that your neighbor calls given that there is an earthquake?
- What is the probability that there is a break-in given that there is an earthquake and the alarm goes off?
- What is the probability that your neighbor calls given that the police call?

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