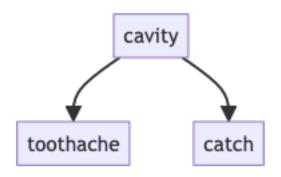
# Bayes' Nets

#### Review

- Efficient encoding of a probabilistic model of a domain
- $\bullet$  Set of nodes, one per variable X
- Directed, acyclic graph
- Conditional distribution for each node
- Implicitly encode joint distributions
  - To see what probability a BN gives to a full assignment, multiply all of the relevant conditionals together:  $\mathbb{P}(x_1, x_2, ..., x_n) = \prod_{i=1}^n \mathbb{P}(x_i | \text{parents}(X_i))$
- eg:



			В		P(B)	
			+b		0.001	
		ф		0.999		
	Λ	_			D(IIA)	
	Α	L	J		P(J A)	
	+a		+j		0.9	
	+a		-j		0.1	
	-a		+j		0.05	
	a		-j		0.95	

$$P(+b)P(-e)P(+a) + 0.001 \times 0.998 \times 0.94$$

P(+b, -e, +a, -j)

 $\mathbb{P}(+\text{cavity}, +\text{catch}, -\text{toothache}) = \mathbb{P}(+\text{cavity})\mathbb{P}(-\text{toothache}|+\text{cavity})\mathbb{P}(+\text{catch}|+\text{cavity}) - \text{eg2:} \qquad 0.001 \times \\ -\text{Joint distribution over } N \text{ variables is } 2^N \text{ rows - Bayes net of } N \text{ nodes with up to } k \text{ parents } \in O(n*2^{k+1})$ 

## Extracting Conditional Independence

- Can use the chain rule and bayes' net reconstitution formula to pull apart factors and determine conditional independence
- Two variables X and Y are conditionally independent given Z iff  $\mathbb{P}(X,Y|Z) = \mathbb{P}(X|Z)\mathbb{P}(Y|Z)$

#### **D-Separation**

• Study independence properties for triples

- Analyze complex cases in terms of member triples
- Question: Are X and Y conditionally independent given evidence variables {Z}?
  Yes, if X and Y "d-separated" by Z
  Consider all (undirected) paths from X to Y
  No active paths = independence!
  A path is active if each triple is active:
  Causal chain A → B → C where B is unobserved (either direction)
  Common cause A ← B → C where B is unobserved
  Common effect (aka v-structure)
   A → B ← C where B or one of its descendents is observed
  All it takes to block a path is a single inactive segment
  Note: These triples are all active (and similar for the other cases, i.e. the variables on either end of the triple can be observed or unobserved)

Figure 1: Screenshot\_2023-10-10\_at\_6.05.43\_PM.png

- - Shaded nodes  $\iff$  observed (taken as a given value)
- To determine conditional independence between  $X_i$  and  $X_j$ , you have to check all possible (UNDIRECTED) paths  $X_i \to X_j$  and see if there is a path s.t. every intermediary node is open
  - The intermediary nodes are triples along the path; open  $\iff$  active

### Topology Limits Distributions

- Adding arcs increases the set of distributions, but has several costs
- Full conditioning can encode any distribution