

Reasoning Under Uncertainty: Belief Networks

CPSC 322 Lecture 26

March 15, 2006

Textbook §9.3

Lecture Overview

Recap

Belief Networks

Belief Network Examples

Conditional independence

- ▶ Sometimes, two random variables might not be entirely independent. However, they can *become* independent after we observe some third variable.
- ▶ Random variable X is **independent** of random variable Y **given** random variable Z if, for all $x_i \in \text{dom}(X)$, $y_j \in \text{dom}(Y)$, $y_k \in \text{dom}(Y)$ and $z_m \in \text{dom}(Z)$,

$$\begin{aligned} P(X = x_i | Y = y_j \wedge Z = z_m) \\ &= P(X = x_i | Y = y_k \wedge Z = z_m) \\ &= P(X = x_i | Z = z_m). \end{aligned}$$

- ▶ That is, knowledge of Y 's value doesn't affect your belief in the value of X , given a value of Z .

Examples of conditional independence

- ▶ The probability that the Canucks will win the Stanley Cup is independent of whether light l_1 is lit given whether there is outside power.
 - ▶ if two random variables are independent, they're **often** also conditionally independent given a third variable.
 - ▶ however, there are special cases where this is not the case¹

¹If you're interested: let C_1 be the proposition that coin 1 is heads; let C_2 be the proposition that coin 2 is heads; let B be the proposition that coin 1 and coin 2 are both either heads or tails. $P(C_1|C_2) = P(C_1)$: C_1 and C_2 are probabilistically independent. But $P(C_1|C_2, B) \neq P(C_1|B)$: if I know both C_2 and B , I know C_1 exactly, but if I only know B I know nothing. Hence C_1 and C_2 are *not* conditionally independent given B .

More examples of conditional independence

- ▶ Whether light l_1 is lit is independent of the position of light switch s_2 given whether there is power in wire w_0 .
 - ▶ two random variables that might not be independent can still be conditionally independent
- ▶ Whether there is someone in a room is independent of whether a light l_2 is lit given the position of switch s_3 .
- ▶ Every other variable may be independent of whether light l_1 is lit given whether there is power in wire w_0 and the status of light l_1 (if it's *ok*, or if not, how it's broken).

Lecture Overview

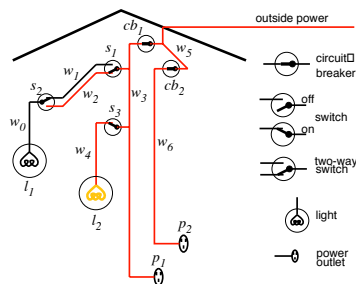
Recap

Belief Networks

Belief Network Examples

Idea of belief networks

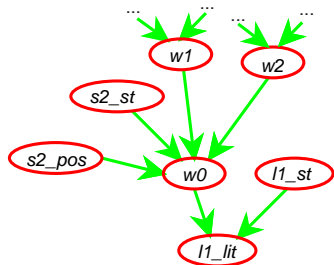
Whether l_1 is lit ($L1_lit$) depends only on the status of the light ($L1_st$) and whether there is power in wire w_0 . Thus, $L1_lit$ is independent of the other variables given $L1_st$ and W_0 . In a belief network, W_0 and $L1_st$ are **parents** of $L1_lit$.



Similarly, W_0 depends only on whether there is power in w_1 , whether there is power in w_2 , the position of switch s_2 ($S2_pos$), and the status of switch s_2 ($S2_st$).

Idea of belief networks

Whether $l1$ is lit ($L1_lit$) depends only on the status of the light ($L1_st$) and whether there is power in wire $w0$. Thus, $L1_lit$ is independent of the other variables given $L1_st$ and $W0$. In a belief network, $W0$ and $L1_st$ are **parents** of $L1_lit$.



Similarly, $W0$ depends only on whether there is power in $w1$, whether there is power in $w2$, the position of switch $s2$ ($S2_pos$), and the status of switch $s2$ ($S2_st$).

Components of a belief network

A belief network consists of:

- ▶ a directed acyclic graph with nodes labeled with random variables
- ▶ a domain for each random variable
- ▶ a set of conditional probability tables for each variable given its parents (which includes prior probabilities for nodes with no parents).

Constructing a belief network

Given a set of random variables, a belief network can be constructed as follows:

- ▶ Totally order the variables of interest: X_1, \dots, X_n
- ▶ Theorem of probability theory (chain rule):
$$P(X_1, \dots, X_n) = \prod_{i=1}^n P(X_i | X_1, \dots, X_{i-1})$$
- ▶ The **parents** pX_i of X_i are those predecessors of X_i that render X_i independent of the other predecessors. That is, $pX_i \subseteq X_1, \dots, X_{i-1}$ and $P(X_i | pX_i) = P(X_i | X_1, \dots, X_{i-1})$
- ▶ So $P(X_1, \dots, X_n) = \prod_{i=1}^n P(X_i | pX_i)$

Lecture Overview

Recap

Belief Networks

Belief Network Examples

Example: Fire Diagnosis

Suppose you want to diagnose whether there is a fire in a building

- ▶ you receive a noisy report about whether everyone is leaving the building.
- ▶ if everyone *is* leaving, this may have been caused by a fire alarm.
- ▶ if there is a fire alarm, it may have been caused by a fire or by tampering
- ▶ if there is a fire, there may be smoke

Example: Fire Diagnosis

First you choose the variables. In this case, all are boolean:

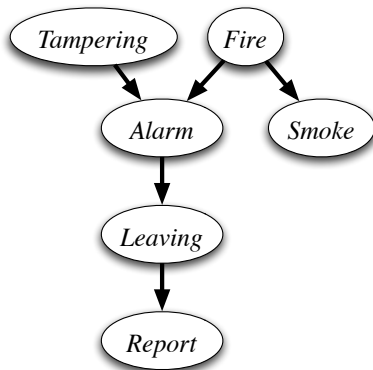
- ▶ **Tampering** is true when the alarm has been tampered with
- ▶ **Fire** is true when there is a fire
- ▶ **Alarm** is true when there is an alarm
- ▶ **Smoke** is true when there is smoke
- ▶ **Leaving** is true if there are lots of people leaving the building
- ▶ **Report** is true if the sensor reports that people are leaving the building

Example: Fire Diagnosis

- ▶ Next, you order the variables: *Fire*; *Tampering*; *Alarm*; *Smoke*; *Leaving*; *Report*.
- ▶ Now evaluate which variables are conditionally independent given their parents:
 - ▶ *Fire* is independent of *Tampering* (given no other information)
 - ▶ *Alarm* depends on both *Fire* and *Tampering*. That is, we are making no independence assumptions about *Fire*, given this variable ordering.
 - ▶ *Smoke* depends only on *Fire*, and is independent of *Tampering* and *Alarm* given whether there is a *Fire*
 - ▶ *Leaving* only depends on *Alarm* and not directly on *Fire* or *Tampering* or *Smoke*. That is, *Leaving* is independent of the other variables given *Alarm*.
 - ▶ *Report* only directly depends on *Leaving*.

Example: Fire Diagnosis

This corresponds to the following belief network:

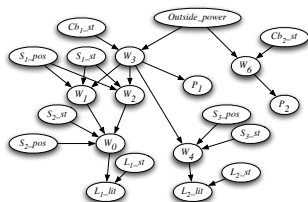
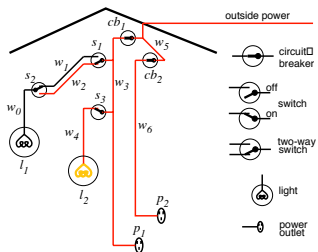


Of course, we're not done until we also come up with conditional probability tables for each node in the graph.

Example: Circuit Diagnosis

The belief network also specifies:

- The domain of the variables:
 $W_0, \dots, W_6 \in \{live, dead\}$
 $S_{1_pos}, S_{2_pos},$ and S_{3_pos} have domain $\{up, down\}$
 S_{1_st} has $\{ok, upside_down, short, intermittent, broken\}$.
- Conditional probabilities, including:
 $P(W_1 = live | s_{1_pos} = up \wedge S_{1_st} = ok \wedge W_3 = live)$
 $P(W_1 = live | s_{1_pos} = up \wedge S_{1_st} = ok \wedge W_3 = dead)$
 $P(S_{1_pos} = up)$
 $P(S_{1_st} = upside_down)$



Example: Circuit Diagnosis

The power network can be used in a number of ways:

- ▶ Conditioning on the status of the switches and circuit breakers, whether there is outside power and the position of the switches, you can simulate the lighting.
- ▶ Given values for the switches, the outside power, and whether the lights are lit, you can determine the posterior probability that each switch or circuit breaker is *ok* or not.
- ▶ Given some switch positions and some outputs and some intermediate values, you can determine the probability of any other variable in the network.

Belief network summary

- ▶ A belief network is a directed acyclic graph (DAG) where nodes are random variables.
 - ▶ A belief network is automatically acyclic by construction.
- ▶ The **parents** of a node n are those variables on which n directly depends.
- ▶ A belief network is a graphical representation of dependence and independence:
 - ▶ **A variable is conditionally independent of its non-descendants given its parents.**