

Reasoning Under Uncertainty: Conditional Independence

CPSC 322 Lecture 25

March 13, 2006

Textbook §9.2 – §9.3

Lecture Overview

Recap

Conditional Independence

Conditioning

- ▶ Probabilistic conditioning specifies how to revise beliefs based on new information.
- ▶ You build a probabilistic model taking all background information into account. This gives the **prior probability**.
- ▶ All other information must be conditioned on.
- ▶ If **evidence** e is all of the information obtained subsequently, the **conditional probability** $P(h|e)$ of h given e is the **posterior probability** of h .

Conditional Probability

The conditional probability of formula h given evidence e is

$$P(h|e) = \frac{P(h \wedge e)}{P(e)}$$

Chain rule:

$$P(f_1 \wedge f_2 \wedge \dots \wedge f_n) = \prod_{i=1}^n P(f_i | f_1 \wedge \dots \wedge f_{i-1})$$

Bayes' theorem:

$$P(h|e) = \frac{P(e|h) \times P(h)}{P(e)}.$$

Probabilistic independence

Random variable X is **independent** of random variable Y if, for all $x_i \in \text{dom}(X)$, $y_j \in \text{dom}(Y)$ and $y_k \in \text{dom}(Y)$,

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

That is, knowledge of Y 's value doesn't affect your belief in the value of X .

Lecture Overview

Recap

Conditional Independence

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 .

Conditional Independence Example

- ▶ Kevin separately phones two students, Alice and Bob.
- ▶ To each, he tells the same number, $n_k \in \{1, \dots, 10\}$.
- ▶ Due to the noise in the phone, Alice and Bob each imperfectly (and independently) draw a conclusion about what number Kevin said.
- ▶ Let the numbers Alice and Bob think they heard be n_a and n_b respectively.
- ▶ Are n_a and n_b probabilistically independent?

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 - ▶ No: we'd expect (e.g.) $P(n_a = 1 | n_b = 1) > P(n_a = 1)$.

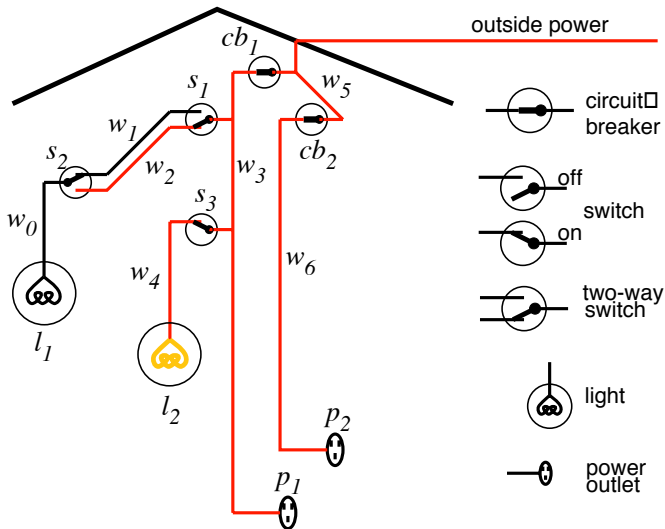
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- ▶ Why are n_a and n_b conditionally independent?

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- ▶ Are n_a and n_b probabilistically independent?
 - ▶ No: we'd expect (e.g.) $P(n_a = 1|n_b = 1) > P(n_a = 1)$.
- ▶ Why are n_a and n_b conditionally independent?
 - ▶ Because if we know the number that Kevin actually said, the two variables are no longer correlated.
 - ▶ e.g., $P(n_a = 1|n_b = 1, n_k = 2) = P(n_a = 1|n_k = 2)$

Example domain (diagnostic assistant)



More 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).