Conditional independence

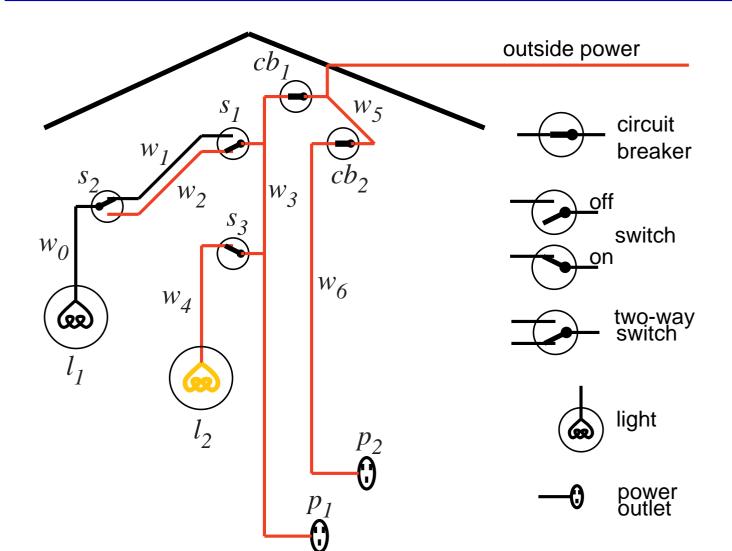
Random variable *X* is independent of random variable *Y* given random variable *Z* if, for all $x_i \in dom(X)$, $y_j \in dom(Y), y_k \in dom(Y)$ and $z_m \in dom(Z)$,

$$P(X = x_i | Y = y_j \land Z = z_m)$$

= $P(X = x_i | Y = y_k \land Z = z_m)$
= $P(X = x_i | Z = z_m).$

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

Example domain (diagnostic assistant)



 Examples of conditional independence

- The identity of the queen of Canada is independent of whether light *l*1 is lit given whether there is outside power.
- > Whether there is someone in a room is independent of whether a light l^2 is lit given the position of switch s^3 .
- > Whether light l1 is lit is independent of the position of light switch s2 given whether there is power in wire w_0 .
- Every other variable may be independent of whether light l1 is lit given whether there is power in wire w₀ and the status of light l1 (if it's ok, or if not, how it's broken).

Idea of belief networks

Whether l1 is lit (l1_lit) depends only on the status of the light (*l*1_*st*) and whether there s2_st is power in wire w0. Thus, l1_lit is independent of the s2_pos [1_st w0 other variables given *l*1_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 ($s2_st$).



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

A belief network is a graph: the nodes are random variables; there is an arc from the parents of each node into that node.

Belief network summary

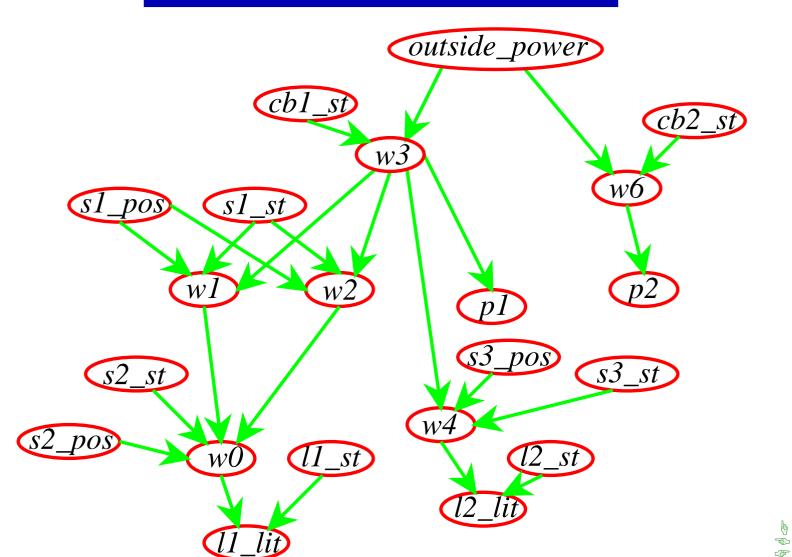
- > A belief network is automatically acyclic by construction.
- A belief network is a directed acyclic graph (DAG) where nodes are random variables.
- The parents of a node *n* are those variables on which *n* directly depends.
- A belief network is a graphical representation of dependence and independence: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

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 (including prior probabilities for nodes with no parents).



Example belief network



Example belief network (continued)

The belief network also specifies:

- The domain of the variables:
 w₀, ..., w₆ have domain {live, dead}
 s₁_pos, s₂_pos, and s₃_pos have domain {up, down}
 s₁_st has {ok, upside_down, short, intermittent, broken}.
- Conditional probabilities, including: $P(w_{1} = live|s_{1}pos = up \land s_{1}st = ok \land w_{3} = live)$ $P(w_{1} = live|s_{1}pos = up \land s_{1}st = ok \land w_{3} = dead)$ $P(s_{1}pos = up)$ $P(s_{1}st = upside_down)$

Constructing belief networks

- To represent a domain in a belief network, you need to consider:
 - > What are the relevant variables?
 - > What values should these variables take?
- What is the relationship between them? This should be expressed in terms of local influence.
- How does the value of one variable depend on the variables that locally influence it (its parents)? This is expressed in terms of the conditional probability tables.

Using belief networks

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.