

Conditional independence

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i.e. for all $x_i \in \text{domain}(X)$, $y_j \in \text{domain}(Y)$, $y_k \in \text{domain}(Y)$ and $z_m \in \text{domain}(Z)$,

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

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

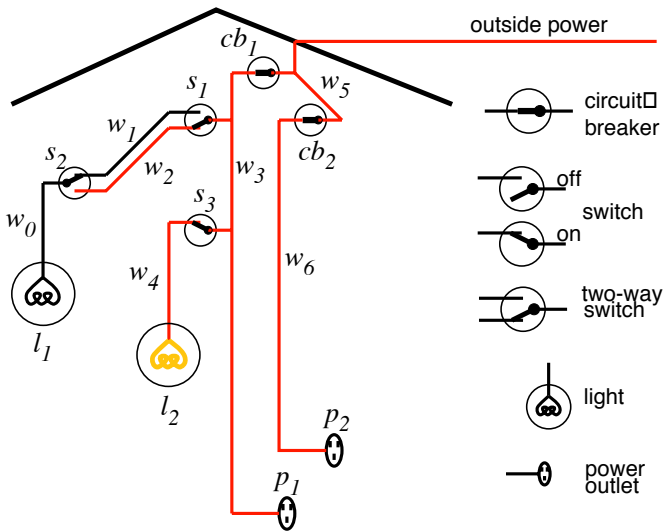
Example

Consider a student writing an exam.

What are reasonable independences among the following?

- Whether the student works hard (W)
- Whether the student is intelligent (I)
- The student's answers on the exam (A)
- The student's mark on an exam (M)

Example domain (diagnostic assistant)



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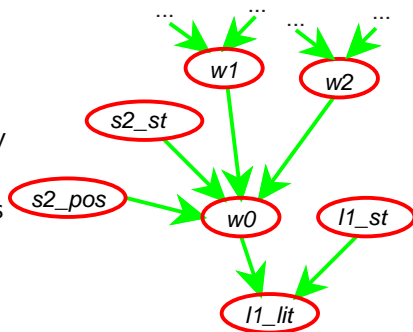
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- Every other variable may be independent of whether light l_1 is lit given whether there is power in wire w_0 and the status of light l_1 (if it's *ok*, or if not, how it's broken).

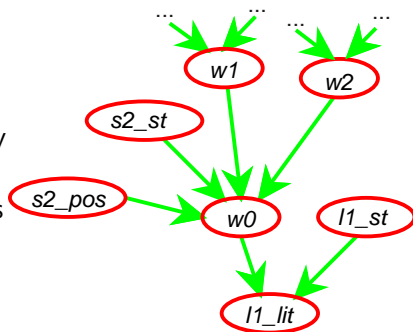
Idea of belief networks

- I_1 is lit ($L1_lit$) depends only on the status of the light ($L1_st$) and whether there is power in wire $w0$.
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- In a belief network, $W0$ and $L1_st$ are **parents** of $L1_lit$.
- $W0$ 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 $s2$ ($S2_st$).



Belief networks

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

Student Writing an Exam Example

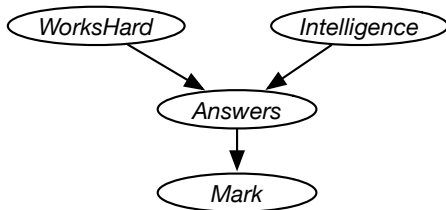
Give a belief network for the variables in order:

- *WorksHard*: Whether the student works hard
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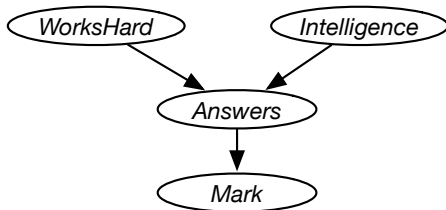
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What if the variables were in the opposite order?

Example: fire alarm belief network

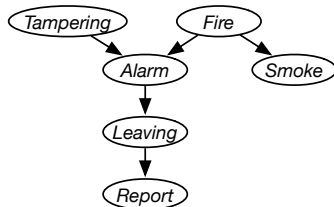
Variables:

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- **Smoke**: what appears to be smoke is coming from a window
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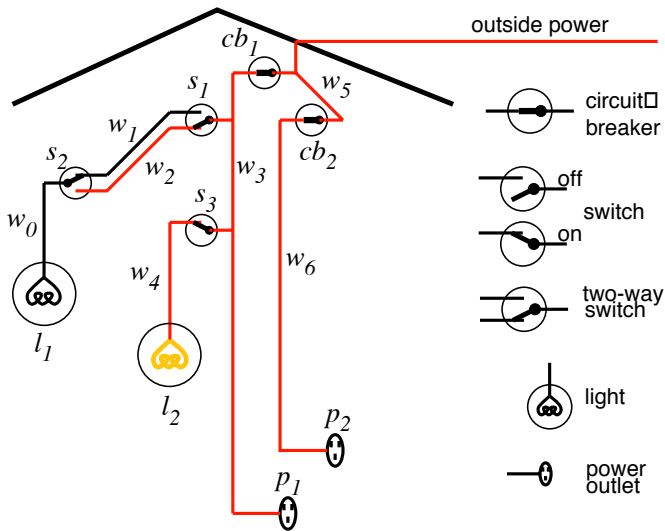
AIPython.org: `bn_report` in `probGraphicalModels.py`

Components of a belief network

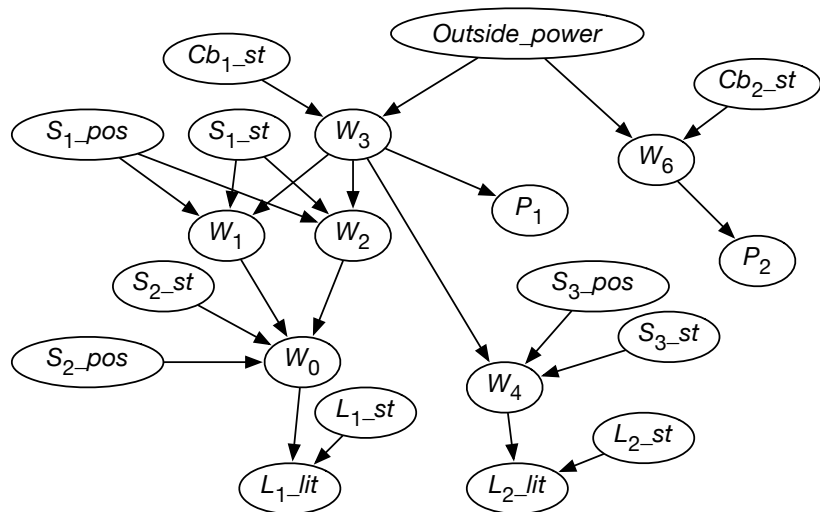
A belief network consists of:

- a directed acyclic graph with nodes labeled with random variables
- a domain for each random variable
- a set of conditional probabilities, one for each variable given its parents (including prior probabilities for nodes with no parents).

Task Domain: Electrical Environment



Example belief network



Example belief network (continued)

The belief network also specifies:

- The domain of the variables:
 - W_0, \dots, W_6 have domain $\{live, dead\}$
 - $S_1_pos, S_2_pos,$ and S_3_pos have domain $\{up, down\}$
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 S_{1_st} has $\{ok, upside_down, short, intermittent, broken\}$.
- Conditional probabilities, including:
 $P(W_1 = live \mid s_{1_pos} = up \wedge S_{1_st} = ok \wedge W_3 = live)$
 $P(W_1 = live \mid s_{1_pos} = up \wedge S_{1_st} = ok \wedge W_3 = dead)$
 $P(S_{1_pos} = up)$
 $P(S_{1_st} = upside_down)$

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- A belief network is automatically acyclic by construction.
- A belief network is a graphical representation of dependence and independence:
 - ▶ A variable is independent of its non-descendants given its parents.

Constructing belief networks

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- How does the value of each variable depend on its parents?
This is expressed in terms of the conditional probabilities.

