

Reasoning under Uncertainty(3): Marginal and Conditional Independence

CPSC 322 – Uncertainty 3
(Giuseppe Carenini)

Textbook – 10.2

What is Co-op?

- Integration of academic studies with relevant, **paid**, and **productive** work experience.
- Co-op students gain skills and experience which prepare them for the future job market and give them **improved employment opportunities** upon graduation.

CPSC Co-op Info Session

Industry work experience
Networking
Money

Think Co-op!

- Date: Nov 8 (Thur)
- Time: 12:30 – 1:30 pm
- Location: DMP 301

Prizes to be won!

Lecture Overview

- Recap with Example
 - Joint Distribution
 - Marginalization
 - Conditional Probability
 - Chain Rule
 - Bayes' Rule
- Marginal Independence
- Conditional Independence

Recap Joint Distribution

➤ 3 binary random variables: $\mathbf{P(H,S,F)}$

- **H** $\text{dom}(\mathbf{H})=\{\mathbf{h}, \neg\mathbf{h}\}$ will have, will not have heart-attack
- **S** $\text{dom}(\mathbf{S})=\{\mathbf{s}, \neg\mathbf{s}\}$ smokes, does not smoke
- **F** $\text{dom}(\mathbf{F})=\{\mathbf{f}, \neg\mathbf{f}\}$ high fat diet, low fat diet

Recap Joint Distribution

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	f		\neg f	
	s	\neg s	s	\neg s
h	.015	.007	.005	.003
\neg h	.21	.51	.07	.18

Recap Marginalization

	f		¬ f	
	s	¬ s	s	¬ s
h	.015	.007	.005	.003
¬ h	.021	.51	.07	.18

$$P(H, S) = \sum_{x \in \text{dom}(F)} P(H, S, F = x)$$

P(H,S)?

P(H)?

P(S)?

Recap Conditional Probability

P(H,S)	s	¬ s	P(H)
h	.02	.01	.03
¬ h	.28	.69	.97
P(S)	.30	.70	

$$P(S | H) = \frac{P(S, H)}{P(H)}$$

P(H S)	s	¬ s
h		
¬ h		

P(S|H)

Recap Chain Rule

$$P(H, S, F) =$$

Recap Bayes Theorem

$$P(S | H) = \frac{P(S, H)}{P(H)}$$

$$P(H | S) = \frac{P(H, S)}{P(S)}$$

$$P(S | H) = \frac{P(H | S)P(S)}{P(H)}$$

Lecture Overview

- Recap with Example
- **Marginal Independence**
- Conditional Independence

Marginal Independence

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

$$P(X = x_i \mid Y = y_k) = P(X = x_i)$$

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

True in our example? E.g. Is Smoking conditional Independent of Heart-attack?

Marginal Independence

True in our example? Is Smoking conditional Independent of Heart-attack?

$P(H,S)$	s	\neg s	$P(H)$
h	.02	.01	.03
\neg h	.28	.69	.97
$P(S)$.30	.70	

$P(S H)$	s	\neg s
h	.666	.334
\neg h	.29	.71

Marginal Independence

Toss two fair coins. Two random variables

- T1 $\text{dom}(T1)=\{h, t\}$
- T2 $\text{dom}(T2)=\{h, t\}$

$P(T1, T2)$	h	t	$P(T1)$
h			
t			

$P(T2)$

$P(T2 T1)$	h	t
h		
t		

Marginal Independence

Consequence...

$P(X, Y) =$

Other Intuitive Examples

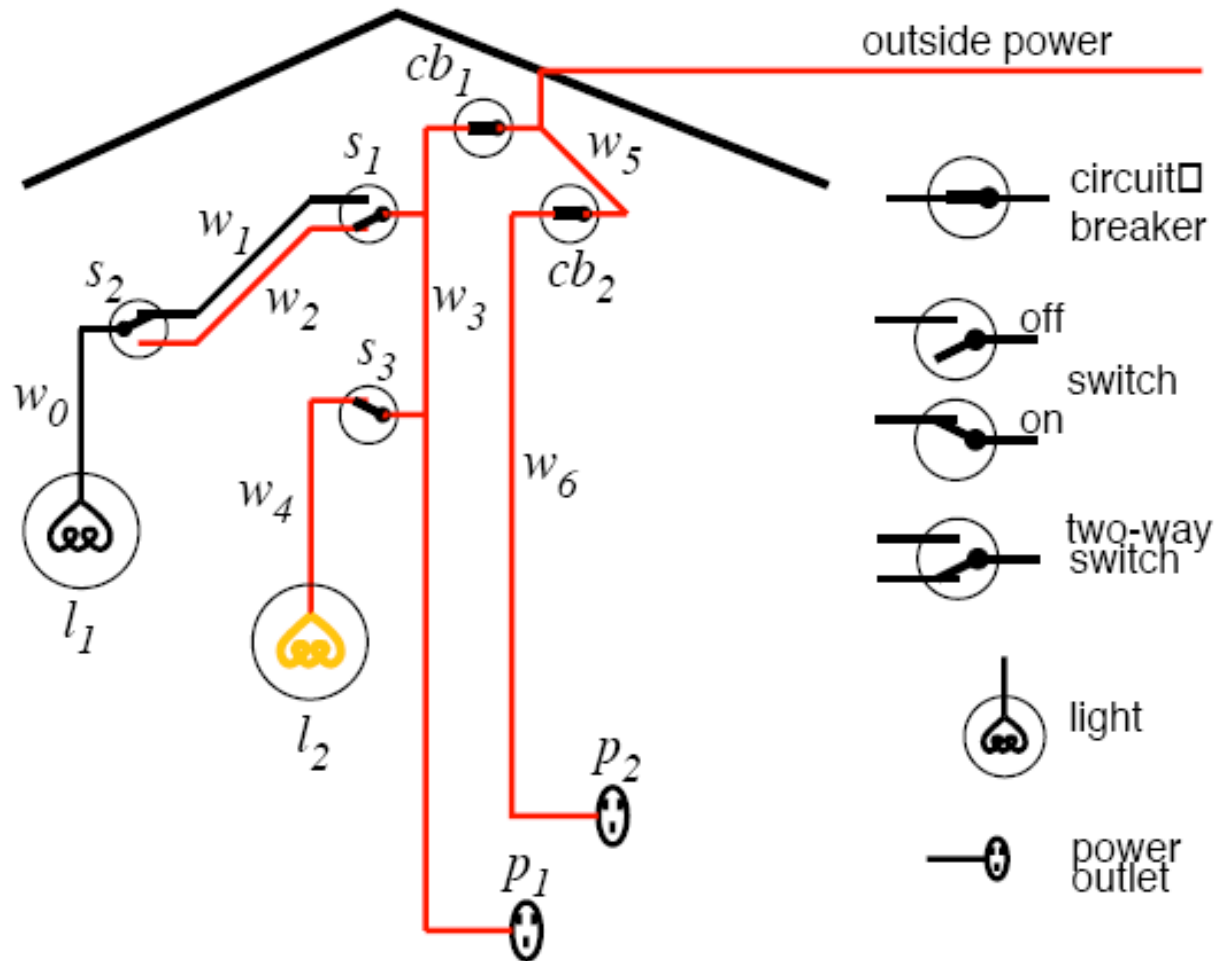
➤ The probability that the Canucks will win the Stanley Cup is independent on

is dependent on....

➤ (Diagnostic assistant) whether light l_1 is lit is not independent of the position of switch s_2

Diagnostic assistant example

- (Diagnostic assistant) whether light l_1 is lit is **not** independent of the position of switch s_2



Lecture Overview

- Recap with Example
- Marginal Independence
- Conditional Independence

Conditional Independence

Sometimes, two variables might not be marginally independent. However, they *become* independent after we observe some third variable

DEF. Random variable **X** is **conditionally independent** of random variable **Y** given random variable **Z** if, for all $x_i \in \text{dom}(X)$, $y_k \in \text{dom}(Y)$, $z_m \in \text{dom}(Z)$

$$P(X= x_i \mid Y= y_k, Z= z_m) = P(X= x_i \mid Z= z_m)$$

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

Conditional Independence Example (1)

- Giuseppe separately phones two students, Alice and Bob.
- To each, he tells the same number, $n_g \in \{1, \dots, 10\}$
- 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 marginally independent?

No, we'd expect (e.g.) $P(n_a = 1 \mid n_b = 1) > P(n_a = 1)$

Conditional Independence

Example (1 cont')

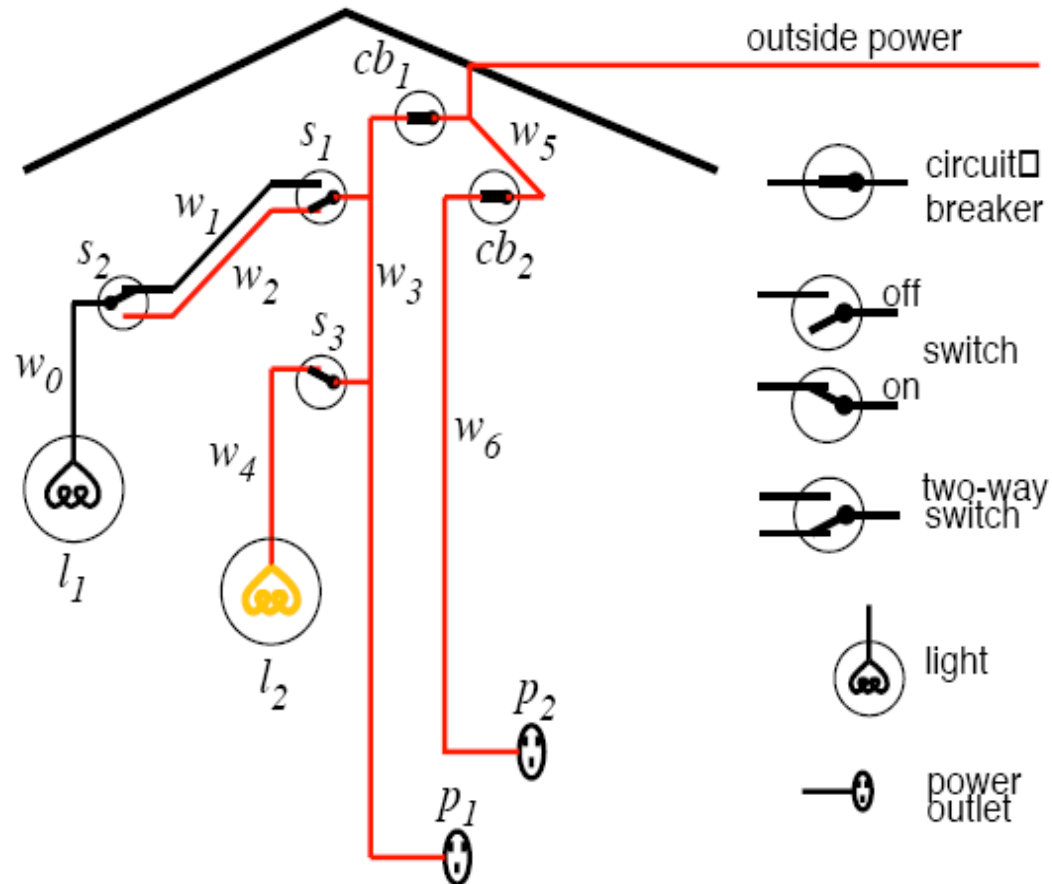
Why are n_a and n_b conditionally independent given n_g ?

Because if we know the number that Giuseppe actually said, the variable n_a is irrelevant

$$\text{e.g. } P(n_a = 1 \mid n_b = 1, n_g = 2) = P(n_a = 1 \mid n_g = 2)$$

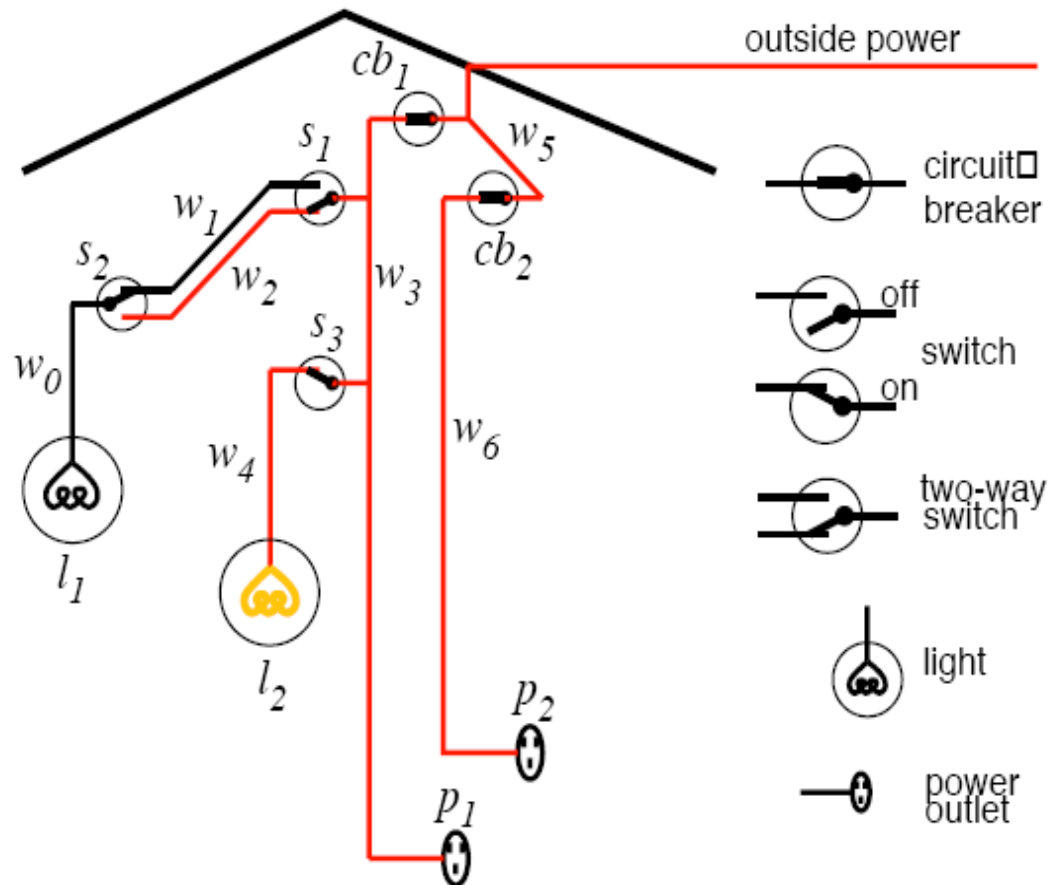
Conditional Independence Example 2

- Whether light l_1 is lit is independent of the position of light switch s_2 given whether there is/isn't power in wire w_0 .



Conditional Independence Example 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).



Next Class

➤ Belief Networks.....