

# Tutorial 2 - Bayesian Networks

COMP9418 – Advanced Topics in Statistical Machine Learning

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**Lecture:** Bayesian Networks

**Topic:** Questions from lecture topics

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## Question 1

Consider the random variables  $X, Y, Z$  which have the following joint distribution:

$Z$	$Y$	$P(Z Y)$
1	1	$a$
0	1	$1-a$
1	0	$b$
0	0	$1-b$

$X$	$P(X)$
1	$a$
0	$1-a$

$X$	$Y$	$P(Y X)$
1	1	$a$
1	0	$1-a$
0	1	$b$
0	0	$1-b$

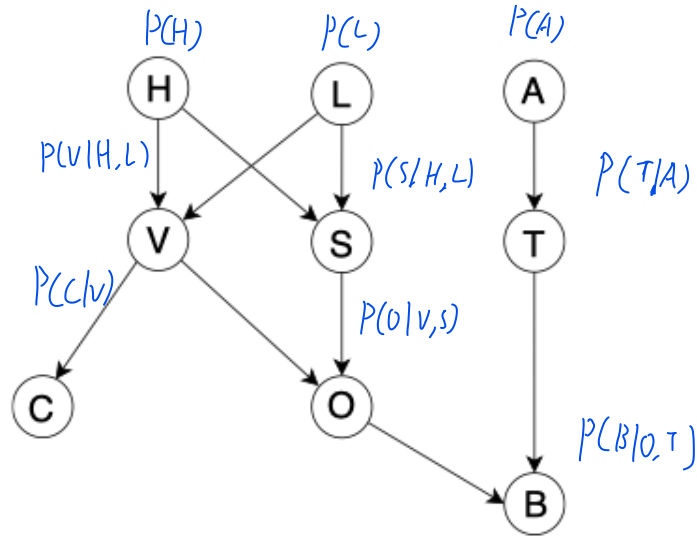
$$P(X, Y, Z) = P(X)P(Y|X)P(Z|Y)$$

- Show that  $X$  and  $Z$  are conditionally independent given  $Y$ .
- If  $X, Y$  and  $Z$  are binary variables, how many parameters are needed to specify a distribution of this form?

$P(X, Y, Z) = P(X)P(Y|X)P(Z|X, Y)$   
 $2^3 - 1 = 7$  don't need the final one since it's just gonna be  
 1 - all the other parameters  
 But we only need  $P(X)$ ,  $P(Y|X)$  and  $P(Z|Y)$ , so we only need 5 parameters

## Question 2

The Bayesian network shown below is a greatly simplified version of a network used for medical diagnosis in an intensive care unit. The diagnostic variables are the hypovolemia ( $H$ ), the left ventricular failure ( $L$ ) and the anaphylaxis ( $A$ ). The intermediate variables are the left ventricular endiastolic volume ( $V$ ), the stroke volume ( $S$ ) and the total peripheral resistance ( $T$ ). The measurement variables are the central venous pressure ( $C$ ), the cardiac output ( $O$ ) and the blood pressure ( $B$ ).  $H, L$  and  $A$  take on the values  $\{true, false\}$ ;  $V, S$  and  $T$  take on the values  $\{low, high\}$ ; and  $C, O$  and  $B$  take on the values  $\{low, medium, high\}$ .



The Bayesian network is fully specified by its CPTs. We have:

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$P(H = \text{true}) = 0.2$	$P(L = \text{true}) = 0.05$	$P(A = \text{true}) = 0.01$
$P(V = \text{low}   H = \text{false}, L = \text{false}) = 0.05$	$P(S = \text{low}   H = \text{false}, L = \text{false}) = 0.05$	
$P(V = \text{low}   H = \text{false}, L = \text{true}) = 0.01$	$P(S = \text{low}   H = \text{false}, L = \text{true}) = 0.95$	
$P(V = \text{low}   H = \text{true}, L = \text{false}) = 0.98$	$P(S = \text{low}   H = \text{true}, L = \text{false}) = 0.5$	
$P(V = \text{low}   H = \text{true}, L = \text{true}) = 0.95$	$P(S = \text{low}   H = \text{true}, L = \text{true}) = 0.98$	
$P(T = \text{low}   A = \text{false}) = 0.3$	$P(T = \text{low}   A = \text{true}) = 0.98$	
$P(C = \text{low}   V = \text{low}) = 0.95$	$P(C = \text{medium}   V = \text{low}) = 0.04$	
$P(C = \text{low}   V = \text{high}) = 0.01$	$P(C = \text{medium}   V = \text{high}) = 0.29$	
$P(O = \text{low}   V = \text{low}, S = \text{low}) = 0.98$	$P(O = \text{medium}   V = \text{low}, S = \text{low}) = 0.01$	
$P(O = \text{low}   V = \text{low}, S = \text{high}) = 0.3$	$P(O = \text{medium}   V = \text{low}, S = \text{high}) = 0.69$	
$P(O = \text{low}   V = \text{high}, S = \text{low}) = 0.8$	$P(O = \text{medium}   V = \text{high}, S = \text{low}) = 0.19$	
$P(O = \text{low}   V = \text{high}, S = \text{high}) = 0.01$	$P(O = \text{medium}   V = \text{high}, S = \text{high}) = 0.01$	
$P(B = \text{low}   O = \text{low}, T = \text{low}) = 0.98$	$P(B = \text{medium}   O = \text{low}, T = \text{low}) = 0.01$	
$P(B = \text{low}   O = \text{low}, T = \text{high}) = 0.3$	$P(B = \text{medium}   O = \text{low}, T = \text{high}) = 0.6$	
$P(B = \text{low}   O = \text{medium}, T = \text{low}) = 0.98$	$P(B = \text{medium}   O = \text{medium}, T = \text{low}) = 0.01$	
$P(B = \text{low}   O = \text{medium}, T = \text{high}) = 0.05$	$P(B = \text{medium}   O = \text{medium}, T = \text{high}) = 0.4$	
$P(B = \text{low}   O = \text{high}, T = \text{low}) = 0.9$	$P(B = \text{medium}   O = \text{high}, T = \text{low}) = 0.09$	
$P(B = \text{low}   O = \text{high}, T = \text{high}) = 0.01$	$P(B = \text{medium}   O = \text{high}, T = \text{high}) = 0.09$	

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- a. Write down the factorised joint distribution defined by the Bayesian network.  $P(H, L, V, S, C, O, A, T, B) = P(H)P(L)P(A)P(V|H,L)P(S|H,L)P(C|V)P(O|V,S)P(T|A)P(B|O,T)$
- b. Show that the above factorised joint distribution is correctly normalised, using the rules of probability.  $\sum_B P(B|O,T) = 1$
- c. Let  $X \perp\!\!\!\perp Y$  denote that  $X$  and  $Y$  are marginally independent and  $X \perp\!\!\!\perp Y|Z$  denote that  $X$  and  $Y$  are conditionally independent given  $Z$ . Using the concept of d-separation, show or refute the following independence statements:
- $H \perp\!\!\!\perp L$  Yes
  - $H \perp\!\!\!\perp A$  Yes
  - $C \perp\!\!\!\perp L$  No
  - $V \perp\!\!\!\perp A|B$  No

- d. For the cases where independence holds in item c, prove these independences using the rules of probability.

$$P(H, L) = P(H)P(L)$$

$$P(H, A) = P(H)P(A) \quad \leftarrow \text{Independence}$$

### Question 3

Jack has three coins,  $C_1$ ,  $C_2$  and  $C_3$ , with  $p_1$ ,  $p_2$  and  $p_3$  as their corresponding probabilities of landing heads. Jack flips coin  $C_1$  twice and then decides, based on the outcome, whether to flip coin  $C_2$  or  $C_3$  next. In particular, if the two  $C_1$  flips come out the same, Jack flips coin  $C_2$  three times next. However, if the  $C_1$  flips come out different, he flips coin  $C_3$  three times next. Given the outcome of Jack's last three flips, we want to know whether his first two flips came out the same.

- Show a Bayesian network structure (graph) for this problem.
- Show the network conditional probability tables (CPTs) for all variables. If you have parameters that are shared among variables, define the CPT once and indicate which variables use that CPT.
- Provide a query that solves this problem.

