

# 1 Three Proofs (26 marks)

- (a) Recall that random variables  $X$  and  $Y$  are conditionally independent, given a third variable  $Z$  if and only if:

$$P(X|Y \wedge Z) = P(X|Z) \quad (1)$$

$$P(Y|X \wedge Z) = P(Y|Z) \quad (2)$$

$$P(X \wedge Y|Z) = P(X|Z)P(Y|Z) \quad (3)$$

Show that Equation 3 follows from Equations 1 and 2. Justify each step of your proof.

Hint: you may find some of the probability rules from Lecture 6 to be useful.

**Marking Scheme:** (6 marks)

- (4 marks) Correct proof
- (2 marks) Each step is justified

**Solutions:**

$P(X|Y \wedge Z) = P(X \wedge Y \wedge Z)/P(Y \wedge Z) = P(X|Z)$  - Conditional Probability from equation 2

$P(X \wedge Y|Z) = P(X \wedge Y \wedge Z)/P(Z)$  - Conditional Probability  
 $= P(X \wedge Y \wedge Z)/P(Y \wedge Z) * P(Y \wedge Z)/P(Z)$  - Alegebra  
 $= P(X|Z) * P(Y \wedge Z)/P(Z)$  - Substitute Equation 2  
 $= P(X|Z)P(Y|Z)$  - Conditional Probability

- (b) The *Markov blanket* of random variable  $X$  in a Bayesian network consists of the set of random variables that are parents, children, or parents of children of  $X$ . For a Bayesian network with random variables  $\mathcal{S}$ , let  $B = \{B_1, B_2, \dots, B_n\} \subset \mathcal{S}$  be the Markov blanket for  $X \in \mathcal{S}$ . See Figure 1 for an example.

Show that  $X \perp\!\!\!\perp Y | B, \forall Y \in \mathcal{S} \setminus B$ . In other words, show that  $X$  is conditionally independent of every other variable in the network, conditioned on its Markov blanket.

Hint: Consider using the fundamental conditional independence relationships discussed in Lecture 8.

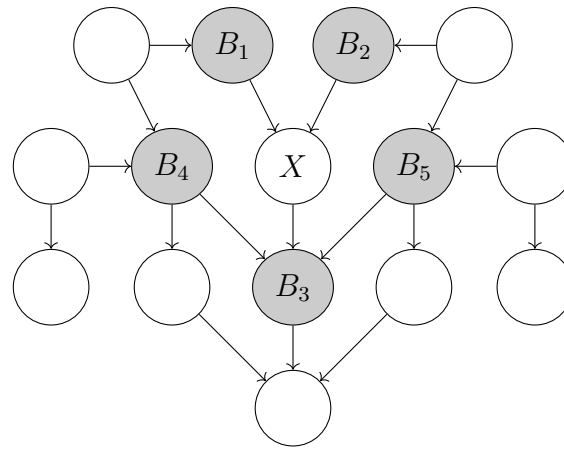


FIGURE 1: An example of a Bayesian network with its Markov blanket shaded grey.

**Marking Scheme:** (10 marks)

- (8 marks) Correct proof
- (2 marks) Proof is succinct and easy to understand

**Solutions:** For  $X$  to be conditionally independent with any node  $N$  given  $B$ , there should not be a path from  $X$  to  $N$  that does not go through  $B$ . Since  $B$  includes all children and parents of  $X$ , this holds for all  $N$  and thus  $X$  is conditionally independent of every variable in the network, conditioned on its Markov blanket

- (c) Not all the random variables in a Bayesian networks are always required to answer a probabilistic query. In fact, all variables that are not ancestors of query variables or evidence variables are irrelevant to the query. Let  $Q = \{Q_1, \dots, Q_m\}$  be the set of query variables and  $e = \{e_1, \dots, e_n\}$  be the set of evidence variables. Prove that the Variable Elimination Algorithm (VEA) returns the same distribution for some query  $P(Q|E)$  if all irrelevant variables are pruned from the Bayesian network.

Hint: try a direct proof. Show that a particular ordering of hidden variable elimination results in the same final distribution returned by the normalization operation.

**Marking Scheme:** (10 marks)

- (8 marks) Correct proof
- (2 marks) Proof is succinct and easy to understand

**Solutions:** TODO