

# KRR - Conditional Independence in Bayesian Networks

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## 1 Conditional Independence and D-separability

This tutorial focuses on the concepts of *conditional independence* and *d-separability* in Bayesian Networks. The definitions below skip a lot of details and do not offer proofs as their role is only to bridge the aforementioned notions.

A Bayesian Network is a probabilistic graphical model that compactly represents a joint distribution over a set of random variables  $\mathcal{X}$ . More precisely, a Bayesian Network is a directed acyclic graph  $\mathcal{G}$  having a node for each random variable in  $\mathcal{X}$ . Each edge in  $\mathcal{G}$  corresponds to a direct causal influence from one variable to another. One of the strengths of this representation is that it captures conditional independence assumptions about the random variables in  $\mathcal{X}$ .

We are interested in evaluating *conditional independence* assertions of the form  $(\mathbf{X} \perp \mathbf{Y} \mid \mathbf{Z})$  where  $\mathbf{X}, \mathbf{Y}, \mathbf{Z}$  are all disjoint subsets of  $\mathcal{X}$ . The semantics of such an expression is *Assuming that  $\mathbf{Z}$  is known would observing  $\mathbf{X}$  bring any information (change our belief) about the variables in  $\mathbf{Y}$ ?*

The conditional independencies in  $\mathcal{G}$  correspond to *d-separation* relations between variables:  $(\mathbf{X} \perp \mathbf{Y} \mid \mathbf{Z}) \iff dsep_{\mathcal{G}}(\mathbf{X}; \mathbf{Y} \mid \mathbf{Z})$ .

We state that  $\mathbf{X}$  and  $\mathbf{Y}$  are d-separated given  $\mathbf{Z}$  if there are no *active paths* between any node  $X \in \mathbf{X}$  and  $Y \in \mathbf{Y}$  given  $\mathbf{Z}$ . In other words,  $\mathbf{X}$  and  $\mathbf{Y}$  are d-separated given  $\mathbf{Z}$  if information does not flow between the two sets of random variables.

A path in  $\mathcal{G}$ :  $X_1 - X_2 - X_3 - \dots - X_N$  is active given  $\mathbf{Z}$  if all its trails of two consecutive edges  $X_{i-1} - X_i - X_{i+1}$  are active given  $\mathbf{Z}$ .

Simple rules can be applied to analyse if a trail  $X_{i-1} - X_i - X_{i+1}$  is active given  $\mathbf{Z}$ :

- Causal trail:  $X_{i-1} \rightarrow X_i \rightarrow X_{i+1}$  is active if  $X_i \notin \mathbf{Z}$  ( $X_i$  is not observed);
- Evidential trail:  $X_{i-1} \leftarrow X_i \leftarrow X_{i+1}$  is active if  $X_i \notin \mathbf{Z}$  ( $X_i$  is not observed);
- Common cause:  $X_{i-1} \leftarrow X_i \rightarrow X_{i+1}$  is active if  $X_i \notin \mathbf{Z}$  ( $X_i$  is not observed);
- Common effect:  $X_{i-1} \rightarrow X_i \leftarrow X_{i+1}$  is active if  $X_i \in \mathbf{Z}$  or any of the descendants  $D \in desc_{\mathcal{G}}(X_i) \in \mathbf{Z}$  ( $X_i$  or any of its descendants is observed);

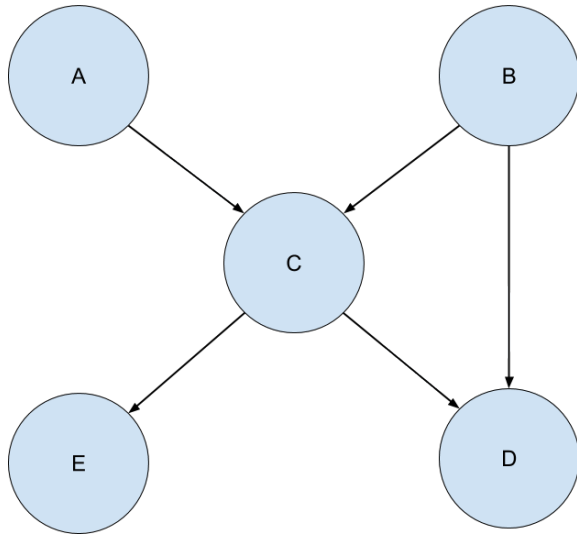
## 2 Tasks

Design an algorithm to check independence assumptions  $(\mathbf{X} \perp \mathbf{Y} \mid \mathbf{Z})$  in a Bayesian Network. A simple idea would be to explore the graph  $\mathcal{G}$  starting from the nodes in  $\mathbf{X}$  following active trails given  $\mathbf{Z}$ . If at any point a node from  $\mathbf{Y}$  is reached, then the independence does not hold. Otherwise, if  $\mathbf{Z}$  blocks all the paths from  $\mathbf{X}$  to  $\mathbf{Y}$ , d-separating the two sets of nodes, then the conditional independence holds.

You are given test files containing the graph structure, various queries, and the correct answers. Such a file has the following structure  $(1 + N + 2M)$  lines):

- two positive numbers on the first line:  $N$  - the number of nodes in  $\mathcal{G}$ , and  $M$  - the number of queries;
- $N$  lines with the name of each variable followed by the names of its parents;
- $M$  lines containing queries expressed as a sequence of names from  $\mathbf{X}$ , the “;” symbol, the names from  $\mathbf{Y}$ , the “|” symbol and then the names of all observed variables from  $\mathbf{Z}$ ;
- $M$  lines containing either “true” or “false” corresponding to the correct answers of the  $M$  independence queries.

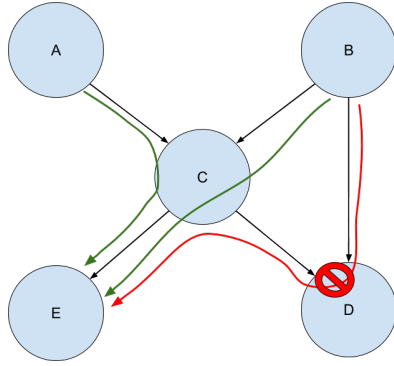
### 3 Example



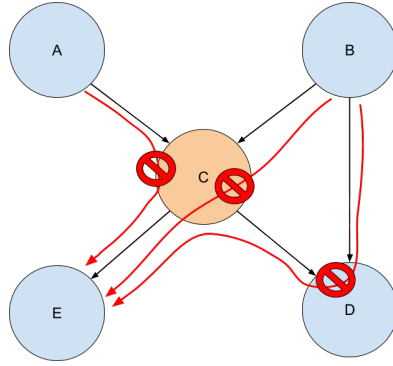
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5 8
A
B
C A B
D C B
E C
A B ; E |
A B ; E | C
E ; D |
E ; D | B
E ; D | C
A ; B |
A ; B | C
A ; B | D
false
true
false
false
true
true
false
false

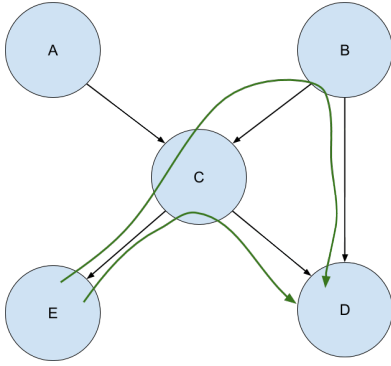
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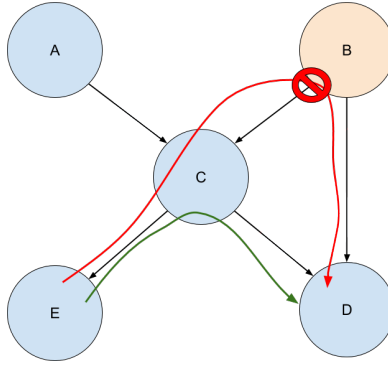
(a)  $(\{A, B\} \not\perp \{E\})$



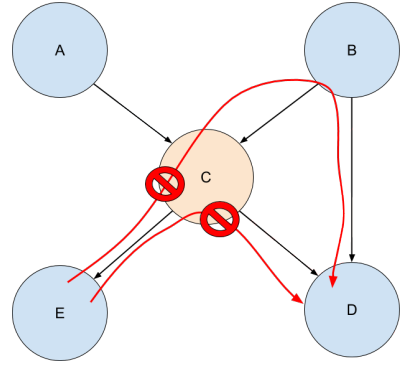
(b)  $(\{A, B\} \perp \{E\} \mid \{C\})$



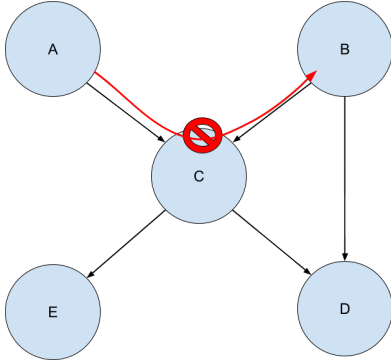
(c)  $(\{E\} \not\perp \{D\})$



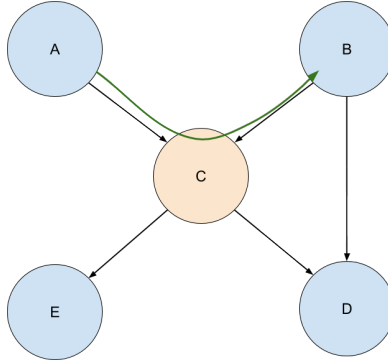
(d)  $(\{E\} \not\perp \{D\} \mid \{B\})$



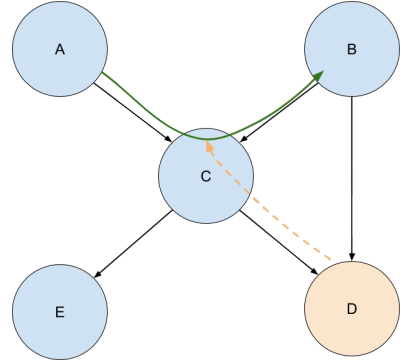
(e)  $(\{E\} \perp \{D\} \mid \{C\})$



(f)  $(\{A\} \perp \{B\})$



(g)  $(\{A\} \not\perp \{B\} \mid \{C\})$



(h)  $(\{A\} \not\perp \{B\} \mid \{D\})$