

# Probabilistic Models: Spring 2013

## Exercise session 3 (exercises 10–13)

**Instructions:** All course participants are requested to submit their exercise solutions as follows:

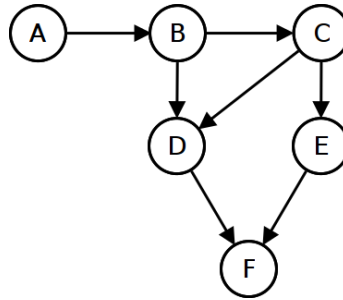
- Deadline: before 12 o'clock noon of the Wednesday when the corresponding exercise session will be held
- Submission as a PDF file by email to joonas.paalasmaa at helsinki.fi (cc: to petri.myllymaki at cs.helsinki.fi)
- If you write the solutions by hand, please scan your paper. However, we strongly recommend that you type your solution by using a word processor. LaTeX is of course especially suitable for typesetting math, and it also has a convenient front-end LyX.
- Use as the title of your paper and subject of the email:  
"ProMo-2013, Exercise session n, yourlastname"
- Use as the file name:  
"ProMo-2013, Exercise session n, yourlastname.pdf"
- In all the exercises, do not just give the answer, but also the derivation of how you obtained it.
- Participants are encouraged to write computer programs to derive solutions whenever appropriate. In this case, please enclose the program source code too as a separate file.

After the exercise session, you are allowed to send a modified version of one of the solutions you sent before the exercise session:

- You can only modify a solution that you submitted before the 12 o'clock deadline.
- You can only send a modified version for one solution, and only if you attended the exercise session.
- Deadline: midnight after the exercise session.
- Submission as before, but this time please send only the modified solution, not the other (unchanged) solutions. Enclose the original solution first and then continue with the new material: first explain what you did wrong in the first time and then continue with modifications.
- As the title of the submission, please use:  
"ProMo-2013, Exercise session n, yourlastname, modified exercise x."

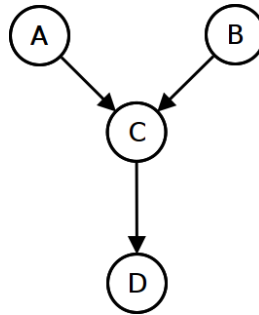
10.

- a) Show by using the d-separation criterion that a node in a Bayesian network is conditionally independent of all the other nodes, given its Markov blanket.
- b) List all d-separations in the following DAG (for each pair of d-separated variables, it is enough to give one set that d-separates them):



11.

- a) List all the marginal and conditional pairwise independencies represented by the following DAG:



- b) Given the above Bayesian network, suppose the variables are binary and the model parameters (conditional probabilities) are as follows:

$$\begin{aligned}
 P(A = 1) &= 0.80 \\
 P(B = 1) &= 0.20 \\
 P(C = 1|A = 0, B = 0) &= 0.30 \\
 P(C = 1|A = 1, B = 0) &= 0.45 \\
 P(C = 1|A = 0, B = 1) &= 0.85 \\
 P(C = 1|A = 1, B = 1) &= 0.65 \\
 P(D = 1|C = 1) &= 0.75 \\
 P(D = 1|C = 0) &= 0.30
 \end{aligned}$$

- i) Construct the joint probability table of the distribution represented by this model.
- ii) What is  $\max P(A, B, C, D)$ ?
- iii) Compute the marginal probabilities  $P(A)$ ,  $P(B)$ ,  $P(C)$  and  $P(D)$ .

iv) Compute the following conditional probabilities:

$$P(A = 1|D = 1)$$

$$P(B = 1|D = 1)$$

$$P(A = 1|D = 0)$$

$$P(B = 1|D = 0)$$

$$P(A = 1|C = 1)$$

$$P(B = 1|C = 1)$$

$$P(A = 0|C = 1)$$

$$P(B = 0|C = 1)$$

$$P(C = 1|D = 0)$$

$$P(C = 1|A = 1)$$

$$P(D = 1|A = 1)$$

$$P(A = 1|C = 1, D = 1)$$

$$P(B = 1|C = 1, D = 1)$$

$$P(A = 1|C = 1, D = 0)$$

$$P(B = 1|C = 1, D = 0)$$

$$P(A = 1|C = 1, B = 1)$$

$$P(A = 1|C = 1, B = 0)$$

$$P(A = 1|B = 1)$$

$$P(A = 1|B = 0).$$

12. List all 3-node Bayesian networks, and partition them into equivalence classes. For the notation, let us call the nodes  $X$ ,  $Y$  and  $Z$ , and let  $XY$  means that there is a directed arc from  $X$  to  $Y$ . For example,  $\{\}$  is an empty network with no arcs, and  $\{XY, XZ, YZ\}$  is a fully connected network with arcs  $X \rightarrow Y$ ,  $X \rightarrow Z$  and  $Y \rightarrow Z$ .

13. [Friedman and Koller 3.18] An edge  $X \rightarrow Y$  in graph  $G$  is said to be *covered* if  $Pa_G(Y) = Pa_G(X) \cup \{X\}$ . Covered edges have the nice property that for every pair of Markov equivalent networks  $G$  and  $G'$ , there exists a sequence of covered edge reversal operations that converts  $G$  to  $G'$ .

- a) Let  $G$  be a graph with a covered edge  $X \rightarrow Y$  and  $G'$  the graph that results by reversing the edge  $X \rightarrow Y$  to produce  $Y \rightarrow X$ , but leaving everything else unchanged. Prove that  $G$  and  $G'$  are Markov equivalent.
- b) Provide a counterexample to this result in the case where  $X \rightarrow Y$  is not a covered edge.