

## Bayesian Networks for Inference

### Exercise T10.1: Graphical models

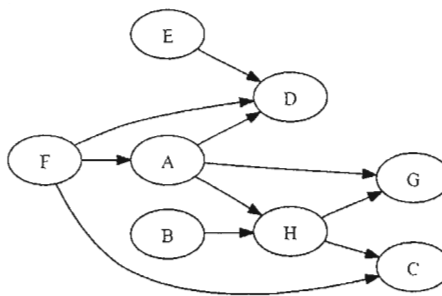
(tutorial)

- (a) Explain the relation between graph structure and corresponding set of random variables.
- (b) Explain the concept of *conditional independence* and how it is exploited in (a).
- (c) What is a *Markov blanket* and how is it used to build a *moral graph*?
- (d) Discuss the *running intersection* property of a *junction tree*.
- (e) Construct a junction tree from a moral graph.

### Exercise H10.1: Directed Acyclic Graphs

(homework, 4 points)

Consider the following DAG:



- (a) (1 point) Give a possible topological sorting of the nodes of this DAG.
- (b) (1 point) The joint distribution of the corresponding  $n$  random variables can be factorized as

$$P(X) = \prod_i^n P(X_i | \text{parents}(X_i)).$$

Write down the factorization for this DAG.

- (c) (1 point) Indicate which nodes belong to the Markov blanket of node A and create the moral graph of the DAG.
- (d) (1 point) Identify the cliques in of the moral graph and construct a junction tree.

**Exercise H10.2: Software****(homework, 2 points)**

Familiarize yourself with software packages for your programming language of choice (e.g., gRain<sup>1</sup> for R or BayesNetToolbox<sup>2</sup> for Matlab or BayesPy<sup>3</sup> for Python).

Implement the “water sprinkler” Bayesian network example in the tutorial by K. Murphy<sup>4</sup>. What is the probability that the sprinkler was active ( $S=\text{true}$ ) after observing that the grass is wet ( $W=\text{true}$ )? What is the probability after the additional observation that it rained recently ( $R=\text{true}$ )?

**Exercise H10.3: Construction of a DAG****(homework, 4 points)**

Consider the binary random variables  $B$  (Burglary),  $E$  (Earthquake),  $A$  (Alarm), and  $R$  (Radio broadcast) which can all take values either “true” (t) or f “false” (f).

Assume our knowledge about their co-occurrence is given by the (conditional) probabilities:  $P(B = t) = 0.01$ ,  $P(E = t) = 10^{-6}$ ,  $P(R = t|E = f) = 0$ ,  $P(R = t|E = t) = 1$ , and

$B$	$E$	$P(A = t B = b, E = e)$
$f$	$f$	0.001
$f$	$t$	0.41
$t$	$f$	0.95
$t$	$t$	0.98

- (1 point) Create a DAG representing the corresponding factorization of the joint distribution.
- (2 points) Implement the DAG with a software package of your choice (see above) and calculate  $P(A)$ ,  $P(A|R = t)$ ,  $P(B|A = t)$  and  $P(B|A = t, R = t)$ .
- (1 point) Explain the phenomenon of *explaining away* at the examples obtained in (b).

**Total 10 points.**

<sup>1</sup> gRain <http://cran.r-project.org/web/views/gR.html>

<sup>2</sup> BayesNetToolbox <http://code.google.com/p/bnt>

<sup>3</sup> BayesPy <http://www.bayespy.org/intro.html> . Another Python package that may be easier to use is <https://github.com/eBay/bayesian-belief-networks> .

<sup>4</sup> K. Murphy. A brief introduction to graphical models and Bayesian networks. <http://people.cs.ubc.ca/~murphyk/Bayes/bnintro.html>