Lecture 9, 10, 11: Junction tree algorithm

- Variable elimination revisited
- Clique trees and junction trees
- Constructing junction trees
- Parametrization
- Junction tree algorithm







We create a clique by connecting all the nodes that are involved

in creating a factor (they would form a clique after elimination)

- The resulting structure is called a *clique tree*
- In general, a clique tree is a singly connected graph in which

nodes are cliques of an underlying graph

Separator sets

A separator set is the intersection of two corresponding cliques

Junction tree property

The cliques containing a particular node form a connected subtree



Constructing a junction tree

- Moralize the graph (if directed)
- Choose a node ordering and find the cliques generated by

variable elimination. This gives a triangulation of the graph

Construct a minimum spanning tree: start with no edges, add

the edges that give maximum cardinality of the separator set

(making sure no cycles are created)

Heuristics for node ordering

- to <u>1</u>. Maximum cardinality: Number the nodes from 1 to n, always assigning the next number to the vertex having the largest set of previously numbered neighbors. Then eliminate nodes from $ar{n}$
- Minimum discrepancy: Always eliminate the node that causes the fewest edges to be added to the induced graph
- Minimum size/weight: Eliminate the node that causes the smallest clique to be created (either in terms of number of nodes, or in terms of number of entries).

Junction tree algorithm

- 1. If the model is directed, moralize it
- 2. Triangulate the undirected graph, using your favorite method
- 3. Parameterize the undirected graph
- 4. Construct the junction tree, using your favorite maximum
- spanning tree algorithm
- 5. Message passing between the nodes!

From the directed to the undirected model



$$\psi(K,C,S) = p(K)p(C|K)p(S|K)$$
$$\psi(C,B) = p(B|C)$$

$$\psi(C, D) = p(D|C)$$
$$\psi(W, C, S) = p(W|S, C)$$



Message passing in the junction tree

- separator sets (i.e., we want all potentials to be proportional to We want to have marginal probabilities in all nodes and marginals over the corresponding variables.
- Let (B, C) be the designated root (we can choose any node)







$$\begin{split} \phi^{**}(C) &= \sum_{B} \psi(B,C) (= \sum_{B} p(B,C) = p(C) \\ \psi^{**}(K,C,S) &= \frac{\phi^{**}(C)}{\phi^{*}(C)} \psi^{*}(K,C,S) \text{(still nothing)} \\ \phi^{**}(C,S) &= \sum_{K} \psi^{**}(K) (= \sum_{K} p(K) p(C|K) p(S|K) = p(C,S)) \\ \psi^{*}(W,C,S) &= \frac{\phi^{**}(C,S)}{\phi^{*}(C,S)} \psi^{*}(W,C,S) (= p(C,S) p(W|C,S) = p(W,C,S)) \end{split}$$

$$\Psi^{**}(K,S,C) \xrightarrow{C} CB\Psi^{*}(B,C) \\ \Phi^{**}(C)$$

Probability propagation

- We introduce the evidence, if any
- Probabilities get propagated using the following equations

(between cliques V and W with separator S):

$$\phi^*(S) = \sum_{V-S} \psi(V)$$
$$\psi^*(W) = \frac{\phi^*(S)}{\phi(S)} \psi(W)$$

These computations do not alter the joint probability distribution

A clique node can send a message to a neighbor after it has

received messages from all other neighbors

This is the same protocol as the sum-product algorithm

Computational complexity

- Constructing the junction tree is done off-line, and is cheap if we are not looking for "optimal" cliques
- On-line, messages are passed on each arc exactly twice
- But the computation here might be expensive!