

**Algorithms for Graphical Models (AGM)**

# **Decomposable models and join forests**

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## **In this lecture**

- Triangulated (decomposable) graphs
- Maximum cardinality search
- Decomposable hypergraphs
- Join forests

## The structure of variable elimination

- In variable elimination, to sum out a variable we have to first multiply all factors containing that variable. This corresponds to merging the associated hyperedges.
- If there is more than one factor this may produce a new hyperedge.
- We then sum out the variable: this corresponds to removing the variable from the (possibly new) hyperedge.
- But it's easier to 'see' this process by looking at the distribution's interaction graph.

## Adding lines to interaction graphs by vertex elimination

- Recall that two variables are connected in the interaction graph if they both appear in a common hyperedge, i.e. they are both variables for some factor.
- Merging all hyperedges containing a given variable  $v$  corresponds to connecting—in the interaction graph—all variables in these hyperedges. A new connection is called a *fill-in*.
- Deleting a vertex from the only hyperedge to contain it corresponds to deleting that vertex from the graph.

## Triangulation by variable elimination

- This process is known as *triangulation*.
- If carried out exactly as described we would end up with an empty graph (since all variables would eventually get deleted).
- Instead we just mark 'eliminated' vertices and treat them *as if* they were deleted.
- It's most easily understood graphically. Cue `triangulation_demo` from `gPy.Examples`

## Complexity of variable elimination

- The triangulated graph gives a nice graphical representation of the complexity of variable elimination with a particular ordering.
- If there are  $n$  variables and  $N$  is the size (number of rows) of the biggest factor, then variable elimination is  $O(nN)$ .
- If this biggest factor has  $\mu$  variables each with no more than  $v$  values then we only get an exponential upper bound:  $N \leq v^\mu$ .
- Need to keep  $\mu$ , the size of the biggest clique, small!

## Looking for good triangulations

- Even deciding whether there is an ordering which bounds the biggest clique by some value is NP-complete. Thus finding an ordering which minimises the biggest clique is a hard problem.
- Nonetheless, there are reasonable heuristic approaches.
- One *greedy* option is, at each point, to eliminate whichever vertex produces the fewest fill-in lines (preferably none).
- Another option is to compute an elimination ordering *backwards*, using *maximum cardinality search*.

## Maximum cardinality search on graphs

Number vertices from  $[n - 1]$  to  $[0]$  in descending order. As the next vertex to number, select the vertex adjacent to the largest number of previously numbered vertices, breaking ties arbitrarily. (Tarjan and Yannakakis, Siam J. Computing, 13:3, 1984)

It's really that simple. Cue `max_card_search_demo.py` from `gPy.Examples`



## Properties of maximum cardinality search

- If there are  $n$  vertices and  $m$  edges in the graph, then maximum cardinality search is  $O(n + m)$ , so it's linear :-).
- If there is a zero fill-in for a graph, maximum cardinality search will find one. :-)
- If there is not, then Kjærulff has shown that the order it finds is generally not that great. :-)

## Key facts about triangulated graphs

- A triangulated graph is *defined* to be an undirected graph such that every cycle of length  $n \geq 4$  has a *chord*: two non-consecutive vertices which are neighbours.
- This is why triangulated graphs are also called *chordal*.
- A graph is triangulated if and only if it has a zero fill-in.

## Some definitions

- An *elimination ordering* for a graph is just an ordering of the vertices of the graph.
- The *fill-in* generated by an elimination ordering is the collection of *extra* lines added by running the vertex elimination algorithm with this ordering.
- An elimination ordering is a *zero fill-in* if its fill-in is empty.

## Towards join forests

- The cliques of a triangulated graph can be arranged in an important data structure: a *join forest*. These are very closely related to the cluster trees we have seen previously.
- A join forest is a collection of trees—connected graphs without cycles—whose vertices are the cliques of the triangulated graphs.
- We will construct them by doing vertex elimination on a distribution's *hypergraph* rather than its interaction graph.

## Decomposable hypergraphs

- A hypergraph is *decomposable* if its reduction is the clique hypergraph of a decomposable (i.e. triangulated) graph.
- This is not (usually) given as the *definition* of decomposable hypergraphs, but it may as well be.
- Note that decomposable hypergraphs are always graphical.
- In the maths literature decomposable hypergraphs are usually called *acyclic hypergraphs*.

## **Graham's algorithm: variable elimination on hypergraphs**

Graham's algorithm: Repeat the following two operations on a hypergraph until neither apply:

1. Delete a vertex that appears in only one hyperedge.
2. Delete a hyperedge that is contained in another hyperedge

A hypergraph is decomposable iff Graham's algorithm deletes all vertices.

## Join trees

A *join tree*  $\mathcal{T}$  for a hypergraph  $\mathcal{H}$  is a tree such that:

1. the vertices of  $\mathcal{T}$  are the hyperedges of  $\mathcal{H}$ , and
2. for all  $h_1, h_2 \in \mathcal{H}$  and any  $h$  on the unique path between  $h_1$  and  $h_2$  in  $\mathcal{T}$ , we have  $h_1 \cap h_2 \subseteq h$ .

The second condition is the *join* or *junction* property. Join trees are often called *junction trees*.

Cue `join_forest_demo` from `gPy.Examples`

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## Join forests

A *join forest* for a hypergraph  $\mathcal{H}$  is simply a collection of join trees for hypergraphs  $\mathcal{H}_i$  where  $\mathcal{H}$  is the disjoint union of the  $\mathcal{H}_i$  such that hyperedges in different trees are disjoint.

Almost always we will deal with join forests containing a single tree.

So, most of the literature talks about junction trees rather than join forests.



## **Decomposable hypergraphs and join forests**

A fundamental theorem is that:

A hypergraph is decomposable if and only if it has a join forest.

## Constructing join forests

- There are number of different ways of constructing a join forest for a decomposable hypergraph.
- Recall that a decomposable hypergraph is always the clique hypergraph of some triangulated graph, and that each triangulated graph has an associated elimination order which is a zero fill-in.
- If we run variable elimination with this ordering then each elimination ‘happens in a clique’. If clique  $C_i$  produces a message which is ‘used’ by clique  $C_j$  then connect these two cliques in the join forest.

## Constructing join forests with Graham's algorithm

1. Delete a vertex that appears in only one hyperedge.
  2. Delete a hyperedge that is contained in another hyperedge
- Graham's algorithm is just 'structural' variable elimination on a decomposable hypergraph using a zero fill-in ordering.
  - When a hyperedge gets deleted just connect it to one of the hyperedges which contained it. Cue `grahams_demo` from `gPy.Examples`