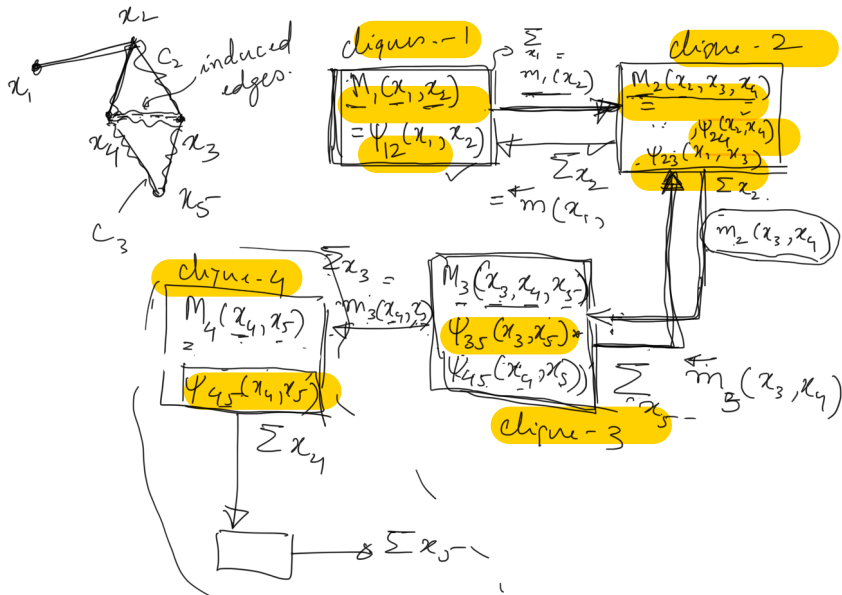


# Computation reuse graph



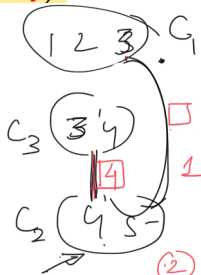
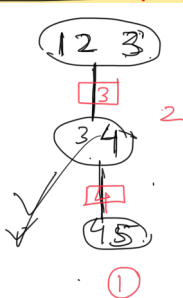
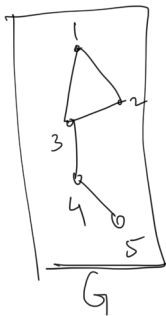
# Junction tree algorithm

- An **optimal** general-purpose algorithm for **exact** marginal/MAP queries
- Simultaneous computation of many queries
- Efficient data structures
- Complexity:  $O(m^w N)$   $w$  = size of the largest clique in (triangulated) graph,  $m$  = number of values of each discrete variable in the clique. → **linear for trees.**
- Basis for many approximate algorithms.
- Many popular inference algorithms special cases of junction trees
  - ▶ Viterbi algorithm of HMMs
  - ▶ Forward-backward algorithm of Kalman filters

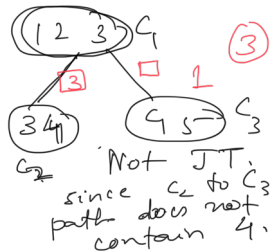
# Junction tree

Junction tree JT of a triangulated graph  $G$  with nodes  $x_1, \dots, x_n$  is a **tree** where

- Nodes = maximal cliques of  $G$
- Edges ensure that if any two nodes contain a variable  $x_i$  then  $x_i$  is present in every node in the unique path between them (**Running intersection property**).



Not JT  
since variable 3 not present in  $C_1$  to  $C_3$  path from



# Constructing a junction tree

Efficient polynomial time algorithms exist for creating a JT from a triangulated graph.

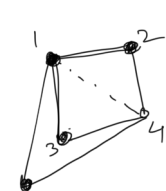
- 1 Enumerate a covering set of cliques
- 2 Connect cliques to get a tree that satisfies the running intersection property.

If graph is non-triangulated, triangulate first using heuristics, optimal triangulation is NP-hard.

Optimal triangulation: A triangulation which gives rise to a JT where the size of the largest clique is smallest.

# Triangulation heuristics

- Choose vertex with smallest degree and connect all its neighbors.
- Choose vertex which will require smallest additional edges to connect neighbors.

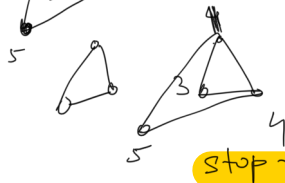


for  $i = 1$  to  $n$

$\pi_i$  = choose the vertex  $i$  for which score is minimum

Connect all neighbors of chosen vertex:  $\pi_i$

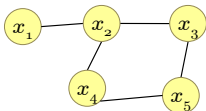
remove  $\pi_i$  from  $G$ .



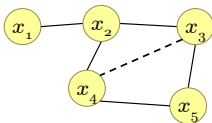
$(1, 3, 4, 5)$   $\rightarrow$  not minimal. yes  
since  $(1, 4)$  is an edge.

# Creating a junction tree from a graphical model

1. Starting graph



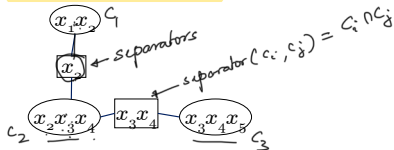
2. Triangulate graph



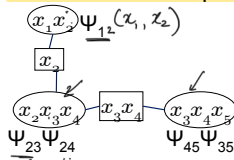
3. Create clique nodes



4. Create tree edges such that variables connected.



5) Assign potentials to exactly one subsumed clique node.



# Finding cliques of a triangulated graph

## Theorem

Every triangulated graph has a **simplicial** vertex, that is, a vertex whose neighbors form a complete set.

Input: Graph  $G$ .  $n$  = number of vertices of  $G$

**for**  $i = 1, \dots, n$  **do**

$\pi_i$  = pick any simplicial vertex in  $G$

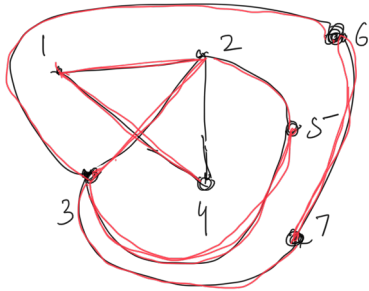
$C_i = \{\pi_i\} \cup \text{Ne}(\pi_i)$

remove  $\pi_i$  from  $G$

**end for**

Return <sup>the</sup> maximal cliques from  $C_1, \dots, C_n$  as nodes of  $\mathcal{J}$

# Example



$$\pi_1 = 1 \quad \pi_2 = 6 \quad \pi_3 = 5$$

$$C_1 = \{1, 2, 4\} \quad C_2 = \{3, 6, 7\} \quad C_3 = \{2, 3, 5\}$$

$$\pi_4 = \text{~~3~~} \quad \pi_5 = 3 \quad \pi_6 = 4$$

$$C_4 = \{3, 7\} \quad \{2, 3\} \quad \{2, 4\}$$





# Connecting cliques to form junction tree

Separator variables = intersection of variables in the two cliques joined by an edge.

## Theorem

*A clique tree that satisfies the running intersection property maximizes the number of separator variables.*

Proof: <https://people.eecs.berkeley.edu/~jordan/courses/281A-fall104/lectures/lec-11-16.pdf>

Input: Cliques:  $C_1, \dots, C_k$

Form a complete weighted graph  $H$  with cliques as nodes and edge weights = size of the intersection of the two cliques it connects.

$T$  = maximum weight spanning tree of  $H$

Return  $T$  as the junction tree.