

EE512A – Advanced Inference in Graphical Models

— Fall Quarter, Lecture 7 —

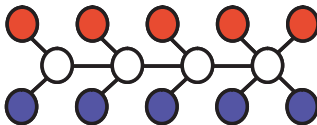
http://j.ee.washington.edu/~bilmes/classes/ee512a_fall_2014/

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Oct 20th, 2014



Announcements

- Reading assignments, posted to our canvas announcements page (<https://canvas.uw.edu/courses/914697/announcements>): `intro.pdf`, `ugms.pdf` on undirected graphical models, and `tree_inference.pdf` on trees.
- Homework 1 is out, due Tuesday (10/21) at 11:45pm, electronically via our assignment dropbox (<https://canvas.uw.edu/courses/914697/assignments>).

Class Road Map - EE512a

- L1 (9/29): Introduction, Families, Semantics
- L2 (10/1): MRFs, elimination, Inference on Trees
- L3 (10/6): Tree inference, message passing, more general queries, non-tree)
- L4 (10/8): Non-trees, perfect elimination, triangulated graphs
- L5 (10/13): triangulated graphs, k -trees, the triangulation process/heuristics
- L6 (10/15): multiple queries, decomposable models, junction trees
- L7 (10/20): junction trees, begin intersection graphs
- L8 (10/22): intersection graphs, inference on junction trees, semirings
- L9 (10/27):
- L10 (10/29):
- L11 (11/3):
- L12 (11/5):
- L13 (11/10):
- L14 (11/12):
- L15 (11/17):
- L16 (11/19):
- L17 (11/24):
- L18 (11/26):
- L19 (12/1):
- L20 (12/3):
- Final Presentations: (12/10):

Finals Week: Dec 8th-12th, 2014.

Decomposition of G and Decomposable graphs

Repeat of both definitions, but on one page.

Definition 7.2.3 (Decomposition of G)

A *decomposition* of a graph $G = (V, E)$ (if it exists) is a partition (A, B, C) of V such that:

- C separates A from B in G .
- C is a clique.

if A and B are both non-empty, then the decomposition is called *proper*.

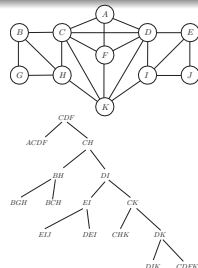
Definition 7.2.4 (Decomposable graph)

A graph $G = (V, E)$ is **decomposable** if either: 1) G is a clique, or 2) G possesses a **proper** decomposition (A, B, C) s.t. both subgraphs $G[A \cup C]$ and $G[B \cup C]$ are decomposable.

Note part 2. It says *possesses*. Bottom of tree might affect top.

Decomposable & numerator/denominator factorization

- Internal nodes in tree are complete graphs that are also separators.
- Decomposable models factor in a useful way.
- With G decomposable, any $p \in \mathcal{F}(G, \mathcal{M}^{(f)})$ can be written as a numerator/denominator of form:



$$p(A, B, C, D, E, F, G, H, I, J, K)$$

$$= \frac{p(A, C, D, F)p(B, C, D, E, F, G, H, I, J, K)}{p(C, D, F)}$$

$$= \frac{p(A, C, D, F)}{p(C, D, F)} \left(\frac{p(B, C, G, H)p(C, D, E, F, H, I, J, K)}{p(C, H)} \right)$$

$$= \dots$$

$$= \frac{p(A, C, D, F)p(B, G, H)p(C, B, H)p(I, E, J)p(E, I, D)p(C, K, H)p(D, K, I)p(D, K, F, C)}{p(C, D, F)p(C, H)p(B, H)p(D, I)p(E, I)p(C, K)p(D, K)}$$

Decomposable models

- When $d(S) > 2$, separator marginal use more than once in the denominator
- The general form of the factorization becomes:

$$p(x) = \frac{\prod_{C \in \mathcal{C}(G)} p(x_C)}{\prod_{S \in \mathcal{S}(G)} p(x_S)^{d(S)-1}} \quad (7.2)$$

where $d(S)$ is the shattering coefficient of separator S .

- Any decomposable model can be written this way
- 4-cycle is not decomposable. Two independence properties that can't be used simultaneously.

$$p(x_1, x_2, x_3, x_4) = \frac{p(x_1, x_2, x_4)p(x_1, x_3, x_4)}{p(x_1, x_4)} = \frac{p(x_1, x_2, x_3)p(x_2, x_3, x_4)}{p(x_2, x_3)} \quad (7.3)$$

Decomposable models

Proposition 7.2.3

All of the maxcliques in a graph lie on the leaf nodes of the binary decomposition tree

Proof.

For a decomposable model, the base case (leaf node) is a clique, otherwise it would not be decomposable. If a leaf was not a maxclique (and only a clique), then that means it is contained in a maxclique, and got split by a separator corresponding to that leaf's parent, but this is impossible since a maxcliques have no separator. □

Proposition 7.2.4

The (nec. unique) set of all minimal separators of graph are included in the non-leaf nodes of the binary decomposition tree. $d(S) - 1$ is the number of times the minimal separator S appears as a given non-leaf node.

Triangulated \equiv decomposable

Theorem 7.2.3

A given graph $G = (V, E)$ is triangulated iff it is decomposable.

Proof.

First, recall from Lemma 4.5.6 that a graph is triangulated iff it is decomposable. To prove the current theorem, we will first show (by induction) that decomposability implies that the graph is triangulated). Next, for the converse, we'll show (also by induction on $n = |V|$) that every minimal separator complete in G implies decomposable.

Tree decomposition (definition)

Definition 7.2.3 (tree decomposition)

Given a graph $G = (V, E)$, a tree-decomposition of a graph is a pair $(\{C_i : i \in I\}, T)$ where $T = (I, F)$ is a tree with node index set I , edge set F , and $\{C_i\}_i$ (one for each $i \in I$) is a collection of subsets of $V(G)$ such that:

- ① $\cup_{i \in I} C_i = V$
- ② for any $(u, v) \in E(G)$, there exists $i \in I$ with $u, v \in C_i$
- ③ for any $v \in V$, the set $\{i \in I : v \in C_i\}$ forms a connected subtree of T

Cluster graphs

Definition 7.2.4 (Cluster graph)

Consider forming a new graph based on G where the new graph has nodes that correspond to clusters in the original G , and has edges existing between two (cluster) nodes only when the corresponding clusters have a non-zero intersection. That is, let $\mathcal{C}(G) = \{C_1, C_2, \dots, C_{|I|}\}$ be a set of $|I|$ clusters of nodes $V(G)$, where $C_i \subseteq V(G), i \in I$. Consider a new graph $\mathcal{J} = (I, \mathcal{E})$ where each node in \mathcal{J} corresponds to a set of nodes in G , and where edge $(i, j) \in \mathcal{E}$ if $C_i \cap C_j \neq \emptyset$. We will also use $S_{ij} = C_i \cap C_j$ as notation.

So two cluster nodes have an edge between them iff there is non-zero intersection between the nodes.

Cluster Trees

If we relax the definition a bit (i.e., drop the requirement for an edge if there exists intersection), and the graph is a tree, then we have what is called a cluster tree.

Definition 7.2.4 (Cluster Tree)

Let $\mathcal{C} = \{C_1, C_2, \dots, C_{|I|}\}$ be a set of node clusters of graph $G = (V, E)$. A cluster tree is a **tree** $\mathcal{T} = (I, \mathcal{E}_T)$ with vertices corresponding to clusters in \mathcal{C} and edges corresponding to pairs of clusters $C_1, C_2 \in \mathcal{C}$. We can label each vertex in $i \in I$ by the set of graph nodes in the corresponding cluster in G , and we label each edge $(i, j) \in \mathcal{E}_T$ by the cluster intersection, i.e., $S_{ij} = C_i \cap C_j$.

Cluster Intersection Property (c.i.p.)

Definition 7.2.4 (Cluster Intersection Property)

We are given a cluster tree $\mathcal{T} = (I, \mathcal{E}_T)$, and let C_1, C_2 be any two clusters in the tree. Then the **cluster intersection property** states that $C_1 \cap C_2 \subseteq C_i$ for all C_i on the (by definition, necessarily) unique path between C_1 and C_2 in the tree \mathcal{T} .

- A given cluster tree might or might not have that property.
- Example on the next few slides.

Running Intersection Property (r.i.p.)

Definition 7.2.4 (Running Intersection Property (r.i.p.))

Let C_1, C_2, \dots, C_ℓ be an ordered sequence of subsets of $V(G)$. Then the ordering obeys the running intersection property (r.i.p.) property if for all $i > 1$, there exists $j < i$ such that $C_i \cap (\cup_{k < i} C_k) = C_i \cap C_j$.

- Cluster j acts as a representative for all of i 's history.
- r.i.p. is defined in terms of clusters of nodes in a graph.
- r.i.p. holds on an (unordered) set of clusters if such an ordering can be found.

Running Intersection Property (r.i.p.)

Given sequence of clusters C_1, C_2, \dots, C_ℓ . Define the **history** (accumulation) of sequence at position i :

$$H_i = C_1 \cup C_2 \cup \dots \cup C_i. \quad (7.2)$$

Innovation (**residual**) or new nodes in C_i not encountered in the previous history, as:

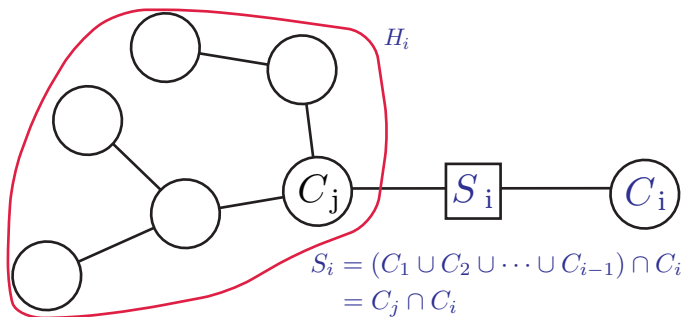
$$R_i = C_i \setminus H_{i-1}. \quad (7.3)$$

Lastly, define the non-innovation, commonality, or **separation** elements between new and previous history:

$$S_i = C_i \cap H_{i-1} \quad (7.4)$$

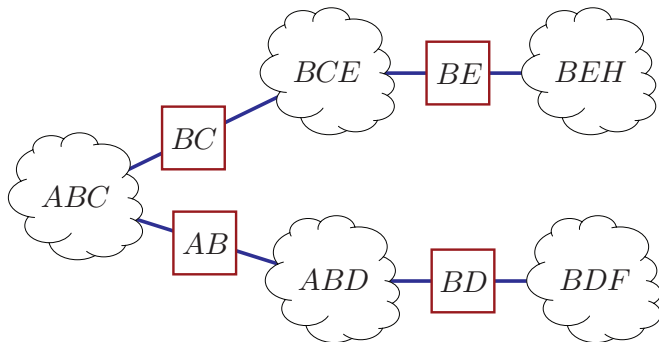
Note $C_i = R_i \cup S_i$, i^{th} cluster consists of the innovation R_i and the commonality S_i .

Running Intersection Property (r.i.p.)



Clusters are in r.i.p. order if the commonality S_i between new and history is fully contained in one element of history. I.e., there exists an $j < i$ such that $S_i \subseteq C_j$.

Example: c.i.p. and r.i.p.



Example of a set of node clusters (within the cloud-like shapes) arranged in a tree that satisfies the r.i.p. and also the cluster intersection property. The intersections between neighboring node clusters are shown in the figure as square boxes. Consider the path or $\{B, E, H\} \cap \{B, D, F\} = \{B\}$.

First Two Properties: c.i.p. \equiv r.i.p

Lemma 7.3.1

The cluster intersection and running intersection properties are identical.

Proof.

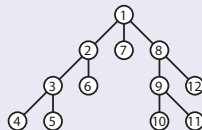
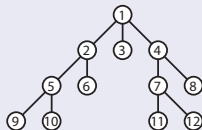
Starting with clusters in r.i.p. order, construct cluster tree by connecting each i to its corresponding j node. This is a tree. Also, take any pair C_k, C_i and assume w.l.o.g. that $k < i$ and hence $C_k \subseteq H_{i-1}$. Then $C_i \cap C_k \subseteq C_i \cap H_{i-1} = S_i \subseteq C_j$. Note that C_j is one node closer to C_k on the path. Repeat this process, but with pair C_k, C_j (if $k < j$) or C_j, C_k (if $j < k$) which decreases the path by one edge, until we get adjacent clusters. This shows c.i.p.

...

First Two Properties: c.i.p. \equiv r.i.p

... proof of Theorem 7.3.1.

Conversely, perform a tree traversal (depth or breadth first search) on cluster tree to produce node ordering.



Then by c.i.p., for any i in that order, and any $k < i$, $C_i \cap C_k \subseteq C_j$ for any j on the unique path between k and i . In particular, $C_i \cap C_k \subseteq C_j$ for $j < i$ being i 's neighbor in the tree. Then $\bigcup_{k < i} (C_i \cap C_k) \subseteq C_j$ implying $C_i \cap \bigcup_{k < i} C_k \subseteq C_j$ and so $C_i \cap \bigcup_{k < i} C_k \subseteq C_i \cap C_j$. On the other hand, we always have that $C_i \cap C_j \subseteq C_i \cap \bigcup_{k < i} C_k$, and the two together give us r.i.p.



Induced sub-tree property (i.s.p.)

Definition 7.3.2 (Induced Sub-tree Property)

Given a cluster tree \mathcal{T} for graph G , the *induced sub-tree property* holds for \mathcal{T} if for all $v \in V$, the set of clusters $C \in \mathcal{C}$ such that $v \in C$ induces a sub-tree $\mathcal{T}(v)$ of \mathcal{T} .

Note, by definition the sub-tree is necessarily connected.

Three properties

Lemma 7.3.3

Induced sub-tree property holds iff cluster intersection property holds

Proof.

Assume induced subtree holds. For any pair C_i, C_j , every $v \in C_i \cap C_j$ induces a sub-tree of \mathcal{T} , and all of these sub-trees overlap on the unique path between C_i and C_j in \mathcal{T} .



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Thus, 1) c.i.p., 2) r.i.p., and 3) the induced sub-tree property are all identical. We'll henceforth refer them collectively as r.i.p.

Tree decomposition and r.i.p.

Recall the definition of **tree decomposition** from the previous lecture, repeated again on the next slide.

Tree decomposition (definition)

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Tree decomposition and r.i.p.

Hence, we see that a tree decomposition (when it exists) is just a cluster tree that satisfies (what we now know to be the) induced sub-tree property (e.g., r.i.p. and c.i.p. as well, i.e., property (3) is r.i.p.), as well as when all nodes and edges are covered (we'll talk more about the notion of "covering" a bit later).

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- We want all original graph (o.g.) clique marginals. Why?

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- Def: **cluster intersection property**, **running intersection property**, **induced sub-tree property**, **r.i.p.**

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- Def: **cluster intersection property**, **running intersection property**, **induced sub-tree property**, r.i.p.
- Next def: **Junction tree**, cluster tree with r.i.p. and edge cover.

Junction Tree

Definition 7.3.4

Given a graph $G = (V, E)$, a **junction tree** corresponding to G (if it exists) is a cluster tree $\mathcal{T} = (\mathcal{C}, E_T)$ having the r.i.p. over the clusters, and where any nodes u, v adjacent via edge $(u, v) \in E(G)$ are together in **at least** one cluster.

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- So, junction tree (JT), for a given graph G , is a cluster tree that: 1) satisfies r.i.p. over the clusters, and 2) includes all edges (edge cover). Not all r.i.p.-satisfying cluster trees need be an edge cover.

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- Clusters in JT need not be original graph cliques!!
- JT could have clusters corresponding to cliques, maxcliques, or neither of the above.

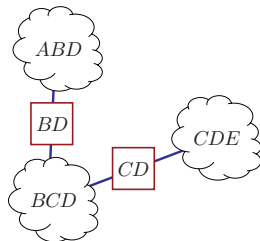
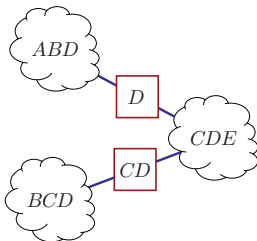
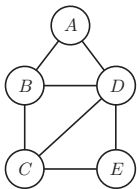
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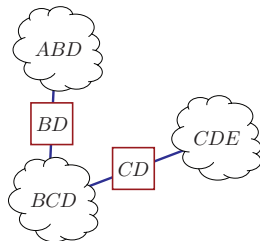
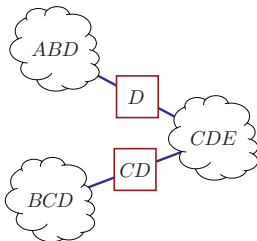
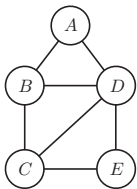
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- Edge cover implies node cover when \exists no isolated nodes.
- Clusters in JT need not be original graph cliques!!
- JT could have clusters corresponding to cliques, maxcliques, or neither of the above.
- If clusters correspond to the original graph cliques (resp. maxcliques) in G , it called a **junction tree of cliques** (resp. maxcliques).

Examples junction trees and not



Questions to answer:

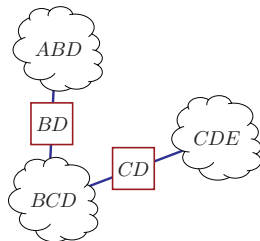
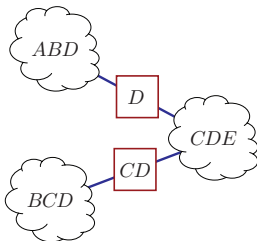
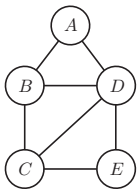
Examples junction trees and not



Questions to answer:

- cluster graph?

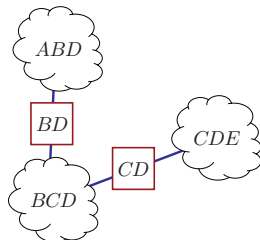
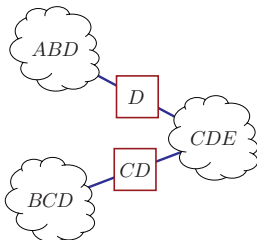
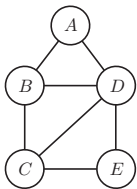
Examples junction trees and not



Questions to answer:

- cluster graph?
- cluster tree?

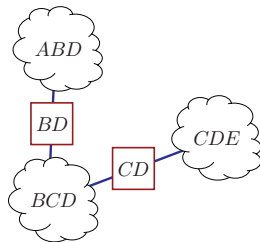
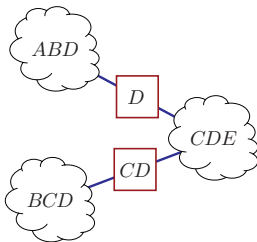
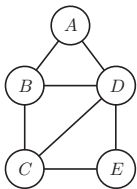
Examples junction trees and not



Questions to answer:

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- Junction tree?

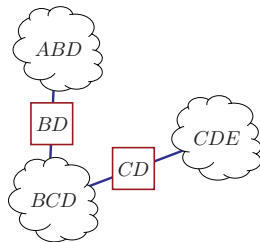
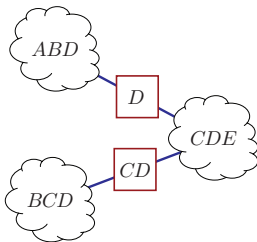
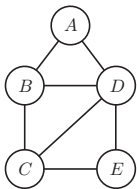
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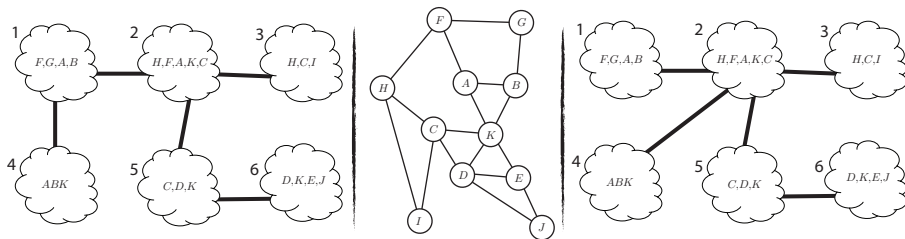
Examples junction trees and not



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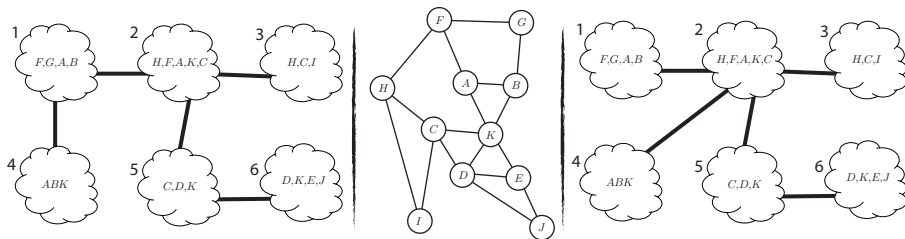
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Examples junction trees and not



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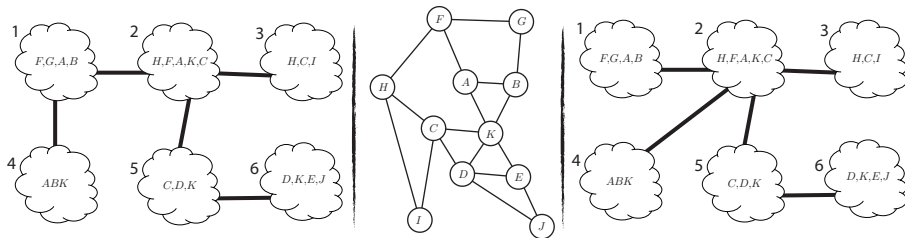
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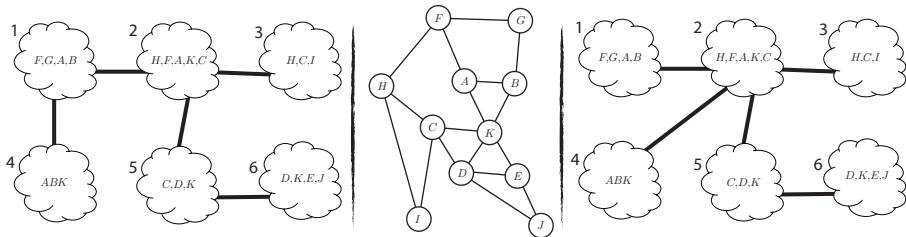
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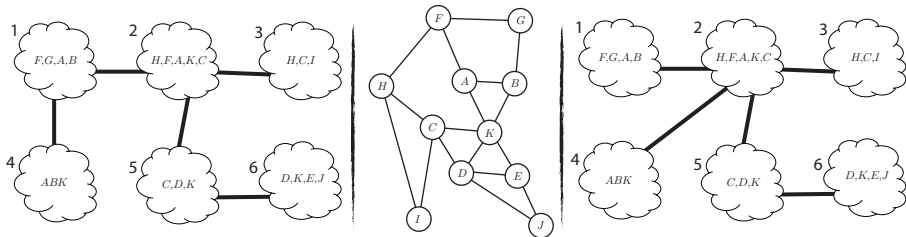
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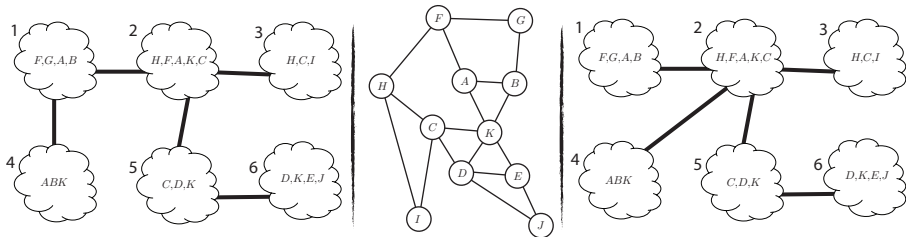
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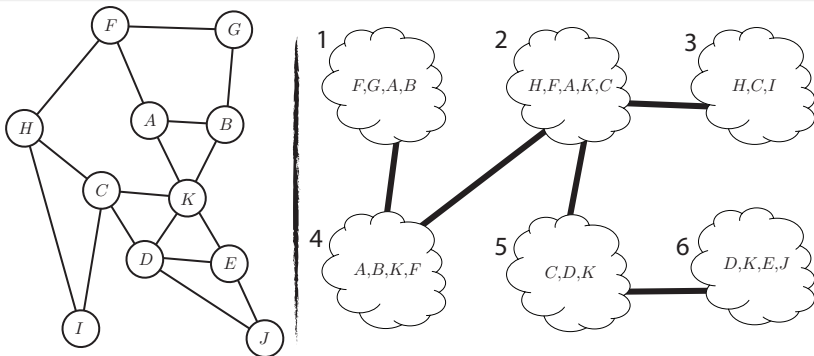
Examples junction trees and not



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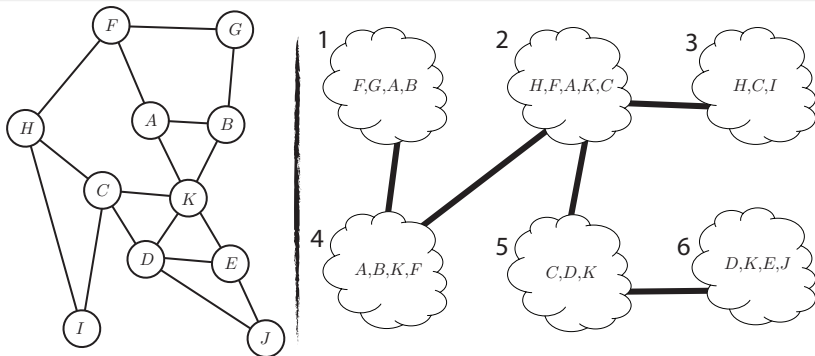
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- Junction tree of maxcliques?

Examples junction trees and not



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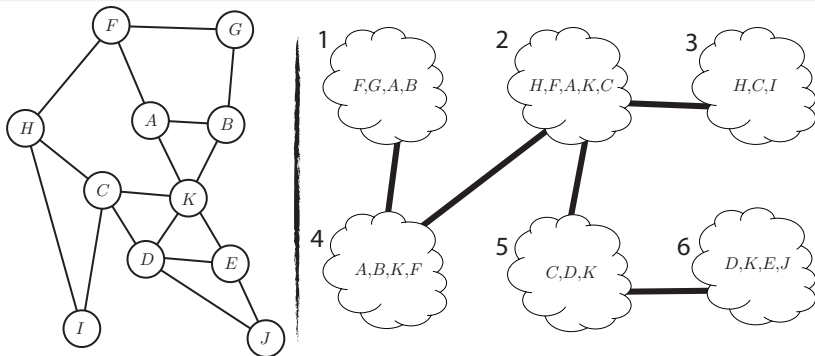
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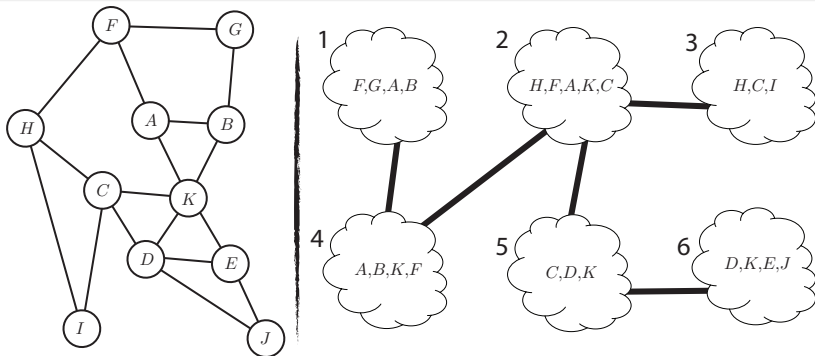
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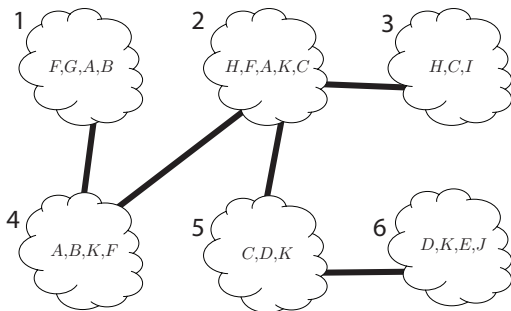
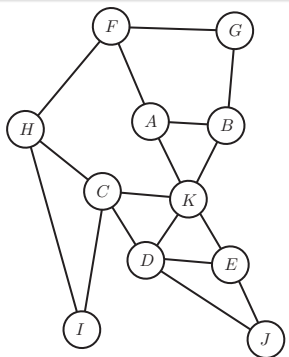
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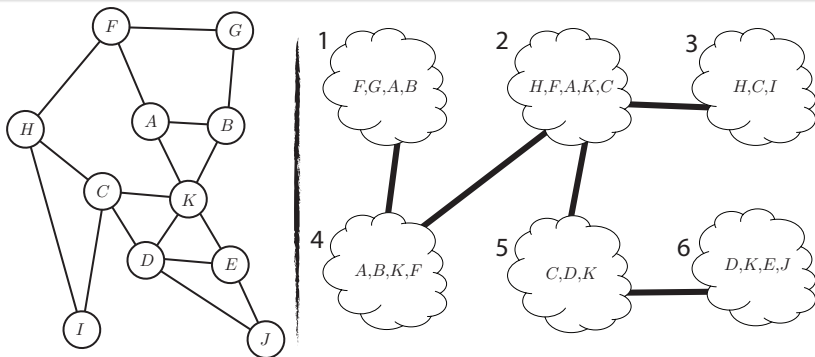
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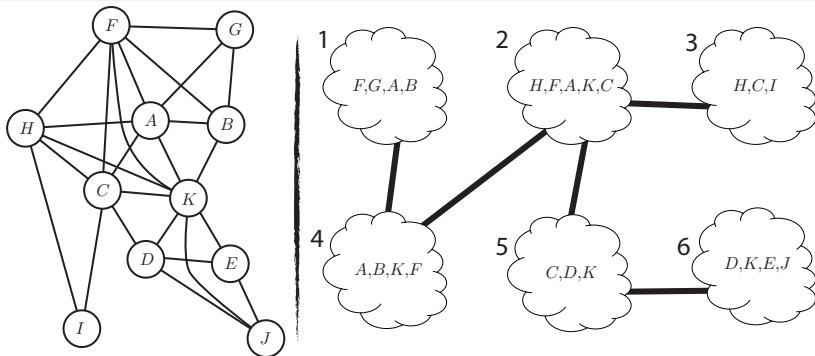
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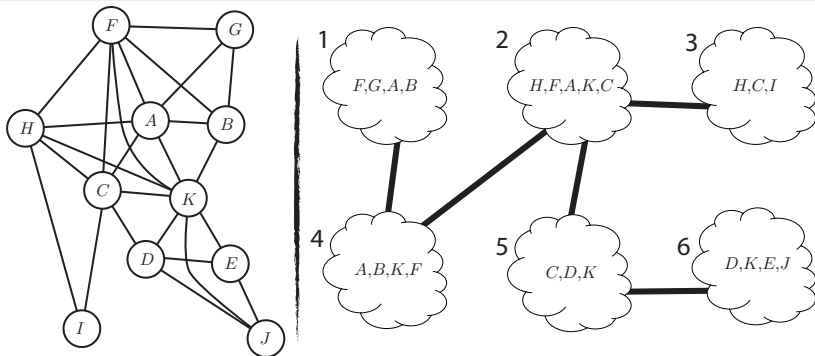
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Examples junction trees and not



Questions to answer:

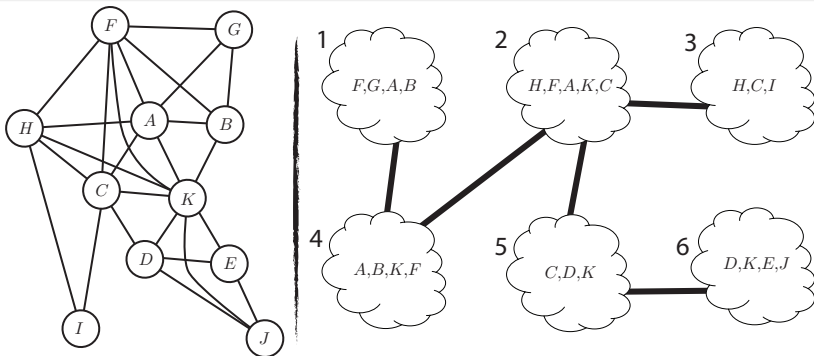
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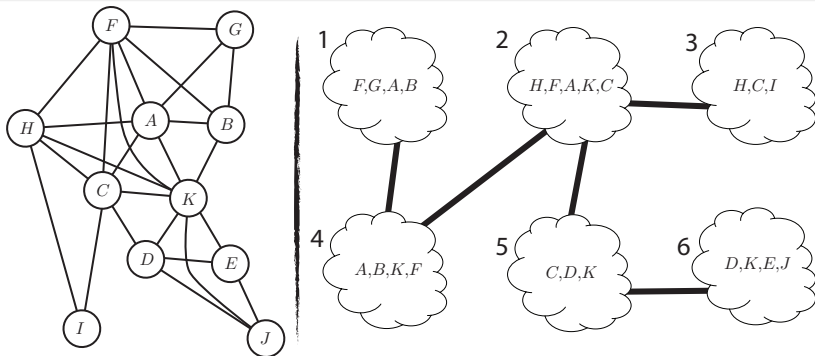
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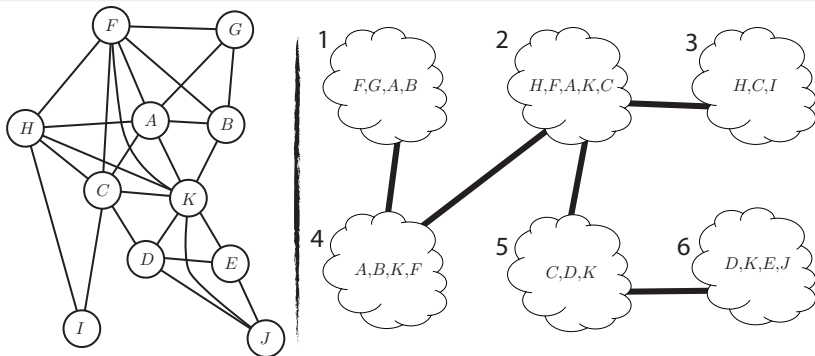
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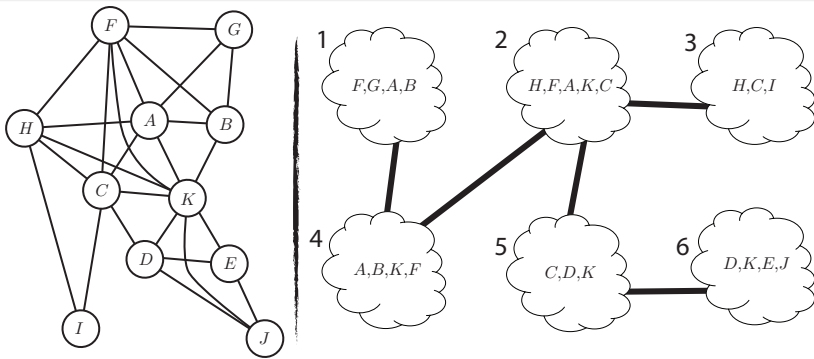
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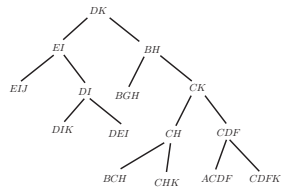
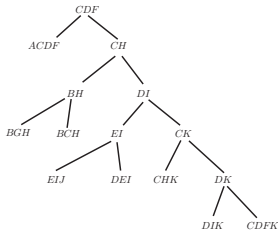
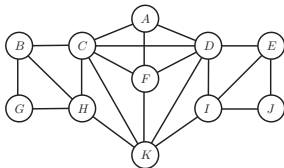
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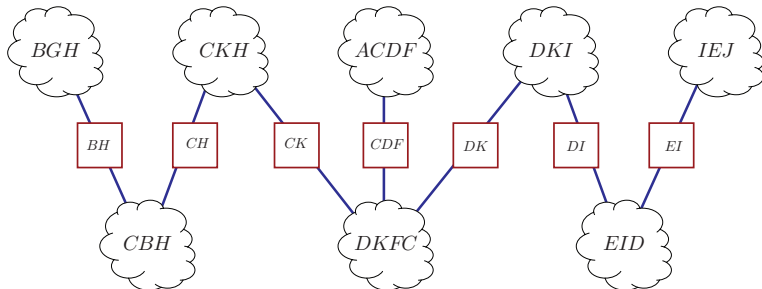
Questions to answer:

- cluster graph?
- cluster tree?
- Junction tree?
- Junction tree of cliques?
- Junction tree of maxcliques?

Examples junction trees and not



Tree of cliques for above graph. Does r.i.p. hold? JT? JT of cliques? JT of maxcliques?



Junction Tree Preserving Operations

Lemma 7.3.5

Given a junction tree, form a new cluster tree as follows. For each cluster C in the JT, choose an order of nodes within C , say c_1, c_2, \dots, c_k , and hang a chain of clusters off of C consisting of $C \setminus \{c_1\}$ hanging from C , $C \setminus \{c_1, c_2\}$ hanging from $C \setminus \{c_1\}$, $C \setminus \{c_1, c_2, c_3\}$ hanging from $C \setminus \{c_1, c_2\}$, and so on. Then the resulting cluster graph is a cluster tree, and moreover it is still junction tree.

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Lemma 7.3.6

Given a junction tree, where (C_i, C_j) are neighboring clusters in the tree, we can merge these two clusters forming a new cluster $C_{ij} = C_i \cup C_j$, and where the neighbors of C_{ij} are the set of neighbors of either C_i or C_j . Then the resulting structure is still junction tree.

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If we keep doing the latter, we'll end up with one complete graph.

Key theorem: JT of maxcliques \equiv triangulated graphs

Theorem 7.3.7

A graph $G = (V, E)$ is decomposable iff a junction tree *of maxcliques* for G exists.

Proof.

a junction tree exists \Leftrightarrow decomposable: Induction on the number of maxcliques. If G has one maxclique, it is both a junction tree and decomposable. Assume true for $\leq k$ maxcliques and show it for $k + 1$.

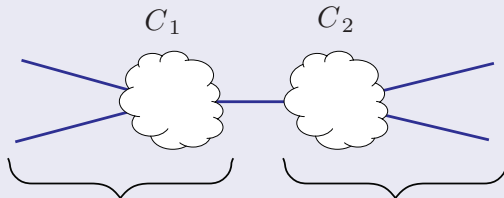
...

Junction tree of maxcliques \equiv triangulated graphs

JT of maxcliques implies Decomposable

... proof continued.

a junction tree exists \Rightarrow decomposable: Let \mathcal{T} be a junction tree of maxcliques \mathcal{C} , and let C_1, C_2 be adjacent in \mathcal{T} . The edge C_1, C_2 in the tree separates \mathcal{T} into two sub-trees \mathcal{T}_1 and \mathcal{T}_2 , with V_i being the nodes in \mathcal{T}_i , $G_i = G[V_i]$ being the subgraph of G corresponding to \mathcal{T}_i , and \mathcal{C}_i being the set of maxcliques in \mathcal{T}_i , for $i = 1, 2$. Thus $V(G) = V_1 \cup V_2$, and $\mathcal{C} = \mathcal{C}_1 \cup \mathcal{C}_2$. Note that $\mathcal{C}_1 \cap \mathcal{C}_2 = \emptyset$. We also let $S = V_1 \cap V_2$ which is the intersection of all the nodes in each of the two trees.



Tree \mathcal{T}_1 with nodes V_1 forming graph $G_1 = G[V_1]$ and max-cliques \mathcal{C}_1 .

Tree \mathcal{T}_2 with nodes V_2 forming graph $G_2 = G[V_2]$ and max-cliques \mathcal{C}_2 .

Junction tree of maxcliques \equiv triangulated graphs

JT of maxcliques implies Decomposable

... proof continued.

Also, the nodes in \mathcal{T}_i are maxcliques in G_i and \mathcal{T}_i is a junction tree for G_i since r.i.p. still holds in the subtrees of a junction tree. Therefore, by induction, G_i is decomposable. To show that G is decomposable, we need to show that: 1) $S = V_1 \cap V_2$ is complete, and 2) that S separates $G[V_1 \setminus S]$ from $G[V_2 \setminus S]$.

If $v \in S$, then for each G_i ($i = 1, 2$), there exists a clique C'_i with $v \in C'_i$, and the path in \mathcal{T} joining C'_1 and C'_2 passes through both C_1 and C_2 .

Because of the r.i.p., we thus have that $v \in C_1$ and $v \in C_2$ and so $v \in C_1 \cap C_2$. This means that $V_1 \cap V_2 \subseteq C_1 \cap C_2$. But $C_i \subseteq V_i$ since C_i is a clique in the corresponding tree \mathcal{T}_i . Therefore

$C_1 \cap C_2 \subseteq V_1 \cap V_2 = S$, so that $S = C_1 \cap C_2$. This means that S contains all nodes that are common among the two subgraphs and moreover that S is complete as desired.

...

Junction tree of maxcliques \equiv triangulated graphs

JT of maxcliques implies Decomposable

... proof continued.

Next, to show that S is a separator, we take $u \in V_1 \setminus S$ and $v \in V_2 \setminus S$ (note that such choices mean $u \notin V_2$ and $v \notin V_1$ due to the commonality property of S). Suppose the contrary that S does not separate V_1 from V_2 , which means there exists a path $u, w_1, w_2, \dots, w_k, v$ for the given u, v with $w_i \notin S$ for all i . Therefore, there is a clique $C \in \mathcal{C}$ containing the set $\{u, w_1\}$. We must have $C \notin \mathcal{C}_2$ since $u \notin V_2$, which means $C \in \mathcal{C}_1$ or $C \subseteq V_1$ implying that $w_1 \in V_1$ and moreover that $w_1 \in V_1 \setminus S$. We repeat this argument with w_1 taking the place of u and w_2 taking the place of w_1 in the path, and so on until we end up with $v \in V_1 \setminus S$ which is a contradiction. Therefore, S must separate V_1 from V_2 . We have thus formed a decomposition of G as $(V_1 \setminus S, V_2 \setminus S, S)$ and since G_i is decomposable (by induction), we have that G is decomposable.

...

Junction tree of maxcliques \equiv triangulated graphs

Decomposable implies JT of maxcliques

... proof continued.

decomposable \Rightarrow a junction tree exists: Since G is decomposable, let (W_1, W_2, S) be a proper decomposition of G into decomposable subsets $G_1 = G[V_1]$ and $G_2 = G[V_2]$ with $V_i = W_i \cup S$. By induction, since G_1 and G_2 are decomposable, there exists a junction tree \mathcal{T}_1 and \mathcal{T}_2 corresponding to maxcliques in G_1 and G_2 . Since this is a decomposition, with separator S , we can form all maxcliques $\mathcal{C} = \mathcal{C}_1 \cup \mathcal{C}_2$ with \mathcal{C}_i maxcliques of V_i for tree \mathcal{T}_i . Choose $C_1 \in \mathcal{C}_1$ and $C_2 \in \mathcal{C}_2$ such that $S \subseteq C_1$ and $S \subseteq C_2$ which is possible since S is complete, and must be contained in some maxclique in both \mathcal{T}_1 and \mathcal{T}_2 . We form a new tree \mathcal{T} by linking $C_1 \in \mathcal{T}_1$ with $C_2 \in \mathcal{T}_2$. We need next to ensure that this new junction tree satisfies r.i.p.

...

Junction tree of maxcliques \equiv triangulated graphs

Decomposable implies JT of maxcliques

... proof continued.

Let $v \in V$. If $v \notin V_2$, then all cliques containing v are in \mathcal{C}_1 and those cliques form a connected tree by the junction tree property since \mathcal{T}_1 is a junction tree. The same is true if $v \notin V_1$. Otherwise, if $v \in S$ (meaning that $v \in V_1 \cap V_2$), then the cliques in \mathcal{C}_i containing v are connected in \mathcal{T}_i including C_i for $i = 1, 2$. But by forming \mathcal{T} by connecting \mathcal{C}_1 and \mathcal{C}_2 , and since v is arbitrary, we have retained the junction tree property. Thus, \mathcal{T} is a junction tree.



Cliques or Maxcliques

Lemma 7.3.8

A junction tree of maxcliques for graph $G = (V, E)$ exists iff a junction tree of cliques for graph $G = (V, E)$ exists.

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- How can we get from one to the other? (Exercise:)

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- How can we get from one to the other? (Exercise:)

Since decomposable is same as triangulated:

Corollary 7.3.9

A graph G is triangulated iff a junction tree of cliques for G exists.

How to build a junction tree

- Maximum cardinality search algorithm can do this. If graph is triangulated, it produces a list of cliques in r.i.p. order.

Maximum Cardinality Search with maxclique order

Algorithm 1: Maximum Cardinality Search: Determines if a graph G is triangulated.

Input: An undirected graph $G = (V, E)$ with $n = |V|$.

Result: is triangulated?, if so MCS ordering $\sigma = (v_1, \dots, v_n)$, and maxcliques in r.i.p. order.

```

1  $L \leftarrow \emptyset$  ;  $i \leftarrow 1$  ;  $\mathcal{C} \leftarrow \emptyset$  ;
2 while  $|V \setminus L| > 0$  do
3   Choose  $v_i \in \operatorname{argmax}_{u \in V \setminus L} |\delta(u) \cap L|$  ; /*  $v_i$ 's previously labeled neighbors has max
   cardinality. */
4    $c_i \leftarrow \delta(v_i) \cap L$  ; /*  $c_i$  is  $v_i$ 's neighbors in the reverse elimination order. */
5   if  $\{v_i\} \cup c_i$  is not complete in  $G$  then
6     return "not triangulated" ;
7   if  $|c_i| \leq |c_{i-1}|$  then
8      $\mathcal{C} \leftarrow (\mathcal{C}, \{c_{i-1} \cup \{v_{i-1}\}\})$  ; /* Append the next maxclique to list  $\mathcal{C}$ . */
9   if  $i = n$  then
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11     $L \leftarrow L \cup \{v_i\}$  ;  $i \leftarrow i + 1$  ;
12 return "triangulated", the ordering  $\sigma$ , and the set of maxcliques  $\mathcal{C}$  which are in r.i.p.
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Maximum Cardinality Search with maxclique order

Algorithm 2: Maximum Cardinality Search: Determines if a graph G is triangulated.

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How to build a junction tree

- Alternatively, we can construct the maxcliques in any form (say by running elimination) and find a maximal spanning tree over the edge-weighted cluster graph, where clusters correspond to maxcliques, and edge weights correspond to the size of the intersection of the two adjacent maxcliques.

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Theorem 7.3.10

A tree of maxcliques \mathcal{T} is a junction tree iff it is a maximum spanning tree on the maxclique graph, with edge weights set according to the cardinality of the separator between the two maxcliques.

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A tree of maxcliques \mathcal{T} is a junction tree iff it is a maximum spanning tree on the maxclique graph, with edge weights set according to the cardinality of the separator between the two maxcliques.

- Note: graph must be triangulated. I.e., maximum spanning tree of a cluster graph where the clusters are maxcliques but the graph is not triangulated will clearly not produce a junction tree.

Other aspects of JTs

- There can be multiple JTs for a given triangulated graph (e.g., consider any graph where $d(S) \geq 3$ for some separator S).

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- first, lets talk a bit about terminology.

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- Going from G' to JT and back to the graph yields the same graph.

Sources for Today's Lecture

- Most of this material comes from the reading handout `tree_inference.pdf`