

Efficient Algorithms for Frequently Asked Questions

4. Hypertree Decompositions

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DaST 
Data • (Systems+Theory)

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University of
Zurich ^{UZH}

<https://lms.uzh.ch/url/RepositoryEntry/17185308706>

FAQ Computation Time Depends on the Structure of the Hypergraph

Consider the following FAQ expression over the Boolean semiring:

$$\Phi() = \bigvee_{(x_1, \dots, x_5) \in \prod_{i \in [5]} \text{Dom}(X_i)} \psi_{12}(x_1, x_2) \wedge \psi_{23}(x_2, x_3) \wedge \psi_{34}(x_3, x_4) \wedge \psi_{15}(x_1, x_5)$$

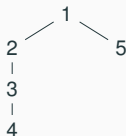
Φ asks whether there is a tuple (x_1, \dots, x_5) such that $\psi_{ij}(x_i, x_j) = \text{true}$

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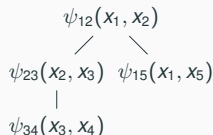
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Hypergraph of Φ , all edges are binary

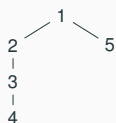


Possible bottom-up evaluation strategy

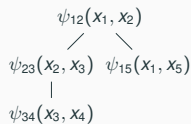
Evaluation strategy known for decades under different names:

- Message passing (in AI literature; Pearl'83)
- Semi-join reduction (in DB literature; Yannakakis'82; discussed in course)

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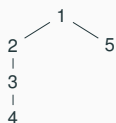


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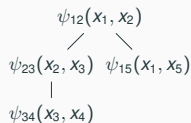


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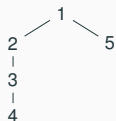


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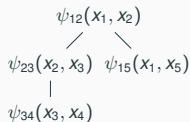
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$$V_{34 \rightarrow 23}(x_3) = \bigvee_{x_4} \psi_{34}(x_3, x_4)$$

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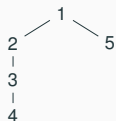
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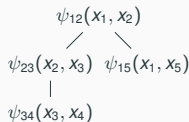
@ ψ_{23} Send up its x_2 -values that are paired with x_3 common to $V_{34 \rightarrow 23}(x_3)$ and ψ_{23} :

$$V_{23 \rightarrow 12}(x_2) = \bigvee_{x_3} \psi_{23}(x_2, x_3) \wedge V_{34 \rightarrow 23}(x_3)$$

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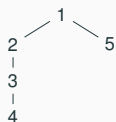
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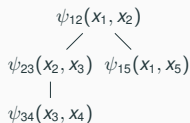
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FAQ Computation Time Depends on the Structure of the Hypergraph



Hypergraph of Φ , all edges are binary



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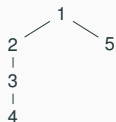
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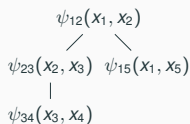
@ ψ_{12} Is there a pair (x_1, x_2) of ψ_{12} with x_1 also in $V_{15 \rightarrow 12}$ and x_2 also in $V_{23 \rightarrow 12}$?

$$\Phi() = \bigvee_{x_1, x_2} \psi_{12}(x_1, x_2) \wedge V_{15 \rightarrow 12}(x_1) \wedge V_{23 \rightarrow 12}(x_2)$$

Computation Time



Hypergraph of Φ , all edges are binary



Possible bottom-up evaluation strategy

All computation steps are local and their cost upper bounded by the factor sizes

- Typical assumption: $|\psi_{ij}| \leq N$ for some value N
- We pass along at most N values between factors
- Local computation is just filtering local values with incoming values
- Overall: linear computation time – This is the best in worst case

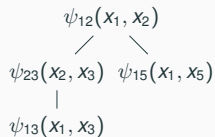
FAQ Computation for a Different Hypergraph

Now, consider a slightly different FAQ Φ' : Same as Φ but $X_4 = X_1$

$$\Phi'() = \bigvee_{x_1, x_2, x_3, x_5} \psi_{12}(x_1, x_2) \wedge \psi_{23}(x_2, x_3) \wedge \psi_{13}(x_1, x_3) \wedge \psi_{15}(x_1, x_5)$$



Hypergraph of Φ' , all edges are binary

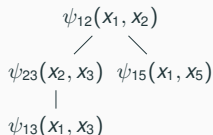
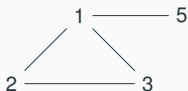


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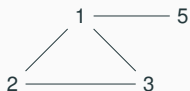
Possible bottom-up evaluation strategy

Computation not anymore local!

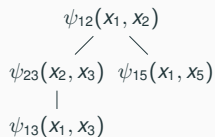
- x_1 needs to be propagated from ψ_{13} through ψ_{23} to ψ_{12}
- ψ_{23} does not have x_1 , so it receives it and forwards it further
- This incurs the cost of carrying x_1 values along two computation steps

$\Rightarrow O(N^2)$ complexity (we will later learn how to do it in $O(N^{1.5})$)

FAQ Computation for a Different Hypergraph



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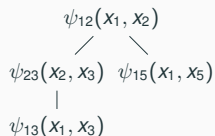


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FAQ Computation for a Different Hypergraph



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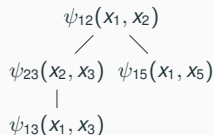


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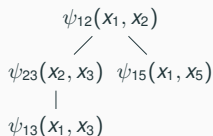
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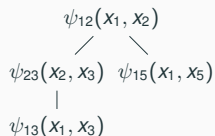
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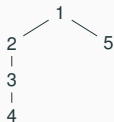
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Why is the Cost of the Second FAQ Higher than of the First One?



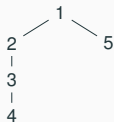
Acyclic Hypergraph



Cyclic Hypergraph

- Left: Only push up information of size $< N$ that is local at factor
- Right: Need to remember longer distance information and push it along
- The difference is reflected in the computational complexity: $O(N)$ vs $O(N^2)$

Why is the Cost of the Second FAQ Higher than of the First One?



Acyclic Hypergraph

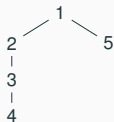


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1. Can we distinguish syntactically the acyclic from the cyclic hypergraphs?

2. Can we “transform” cyclic hypergraphs into acyclic ones?

Answer to Question 1: Acyclic Hypergraphs (Overview)

Several acyclicity notions exist. Studied in this course: α -acyclic & β -acyclic

FAQs **without free variables** can be computed in:

- **Linear time in the size of input factors if its hypergraph is α -acyclic**

Assumption: Each factor ψ_S represented as list of tuples \mathbf{x}_S with $\psi_S(\mathbf{x}_S) \neq \mathbf{0}$

- **Linear time in the size of input factors if its hypergraph is β -acyclic**

Assumption: Each factor represented compactly as box, e.g., for (#)SAT

FAQs **with free variables**:

- *In principle* as above, BUT hypergraph is also **free-connex**
- Linear time for precomputation
- Then output the answer in constant time per tuple (enumeration delay)
- \Rightarrow **Linear time in input size plus output size**

Answer to Question 2: Hypertree Decompositions (Overview)

Hypertree decompositions

- Transform an arbitrary hypergraph into a **hypertree**
- Measure of how close the hypergraph is to a hypertree: **width**
- Complexity of transformation is $\mathcal{O}(N^w)$, where
 - N is the maximal size of an input factor
 - w is the width of the hypergraph
- Once we have a hypertree \rightarrow see answer to Question 1

Hypertree Decompositions

Hypertree Decompositions: Definition

A hypertree decomposition \mathcal{T} of a hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$ is a pair (T, χ) with:

- T is a tree
- χ is a function mapping each node in T to a subset of \mathcal{V} called *bag*

Properties of a decomposition $\mathcal{T} = (T, \chi)$:

- **Coverage:** $\forall e \in \mathcal{E}$, there is a node $t \in T$ such that $e \subseteq \chi(t)$
- **Connectivity:** $\forall v \in \mathcal{V} : \{t \mid t \in T, v \in \chi(t)\}$ forms a connected subtree in T

Observation: Each node $t \in T$ of the hypertree decomposition \mathcal{T} represents the sub-hypergraph \mathcal{H}' of \mathcal{H} induced by the nodes $\chi(t)$ of \mathcal{H}

- The nodes of \mathcal{H}' are $\chi(t)$
- The hyperedges of \mathcal{H}' are \mathcal{H} 's hyperedges restricted to the nodes $\chi(t)$ of \mathcal{H}

Hypertree Decompositions for the Triangle Hypergraph

Triangle query: $\Phi(x_1, x_2, x_3) = \psi_{12}(x_1, x_2) \otimes \psi_{23}(x_2, x_3) \otimes \psi_{13}(x_1, x_3)$



Task: Construct a hypertree decomposition
with one bag per edge

Hypertree Decompositions for the Triangle Hypergraph

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Task: Construct a hypertree decomposition with one bag per edge

First trial



1 not included in the middle bag
Connectivity violated!

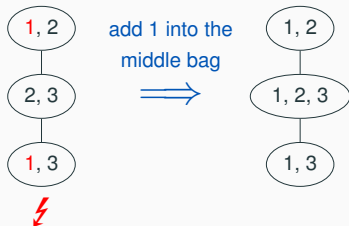
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First trial



add 1 into the
middle bag



remove
redundant bags



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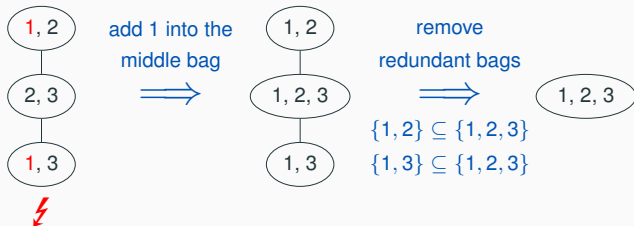
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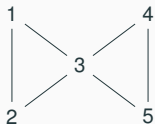


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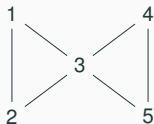
There are five other possibilities. All violate the connectivity condition.

The only hypertree decomposition without redundant bags has only one bag

Hypertree Decompositions for the Bowtie Hypergraph



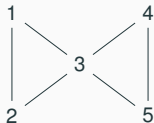
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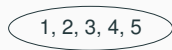
Possible hypertree decompositions

1, 2, 3, 4, 5

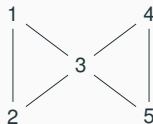
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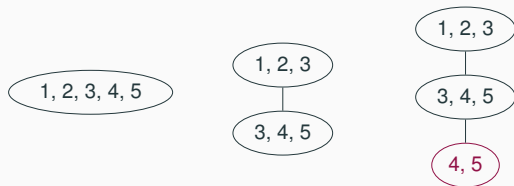
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Hypertree Decompositions for the Bowtie Hypergraph



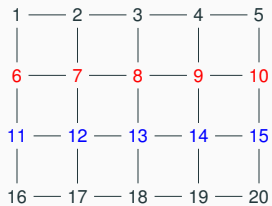
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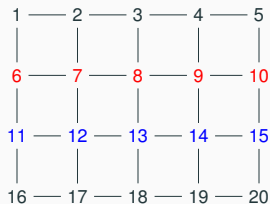
The bag $\{4, 5\}$ is redundant since it is included in bag $\{3, 4, 5\}$.

Redundant bags need not be considered as they add no extra information.

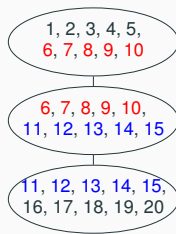
Hypertree Decompositions for the Grid Hypergraph



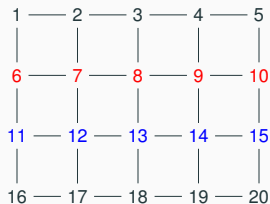
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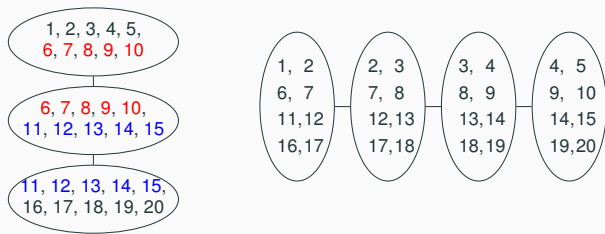
Possible hypertree decompositions



Hypertree Decompositions for the Grid Hypergraph



Possible hypertree decompositions



Hypertree Decompositions for 4-Cycle and 6-Cycle Hypergraphs

4-Cycle



Hypertree decompositions

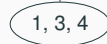
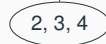
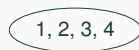


Hypertree Decompositions for 4-Cycle and 6-Cycle Hypergraphs

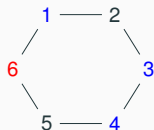
4-Cycle



Hypertree decompositions



6-Cycle

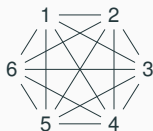


Hypertree Decompositions for Clique and Loomis-Whitney Hypergraphs

Clique of degree n :

Hypergraph $([n], \binom{[n]}{2})$

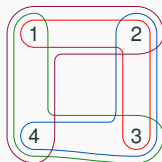
Clique-6



Loomis-Whitney of degree n :

Hypergraph $([n], \binom{[n]}{n-1})$

Loomis-Whitney-4

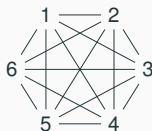


Hypertree Decompositions for Clique and Loomis-Whitney Hypergraphs

Clique of degree n :

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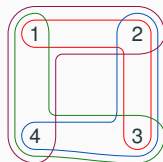
Clique-6



Loomis-Whitney of degree n :

Hypergraph $([n], \binom{[n]}{n-1})$

Loomis-Whitney-4



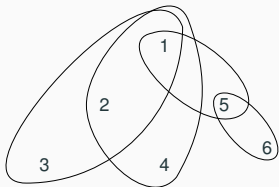
Both hypergraphs only admit the trivial decomposition with one bag

Join Trees: Hypertree Decompositions with One Bag per Hyperedge

α -acyclic hypergraphs admit hypertree decompositions with one bag per hyperedge

- Best decompositions, as no merging of factors in a bag is necessary
- Such decompositions are called **Join Trees**
- Hypergraphs are α -acyclic precisely when they admit join trees

Hypergraph

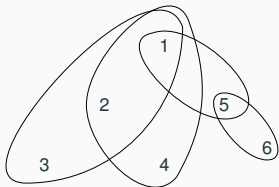


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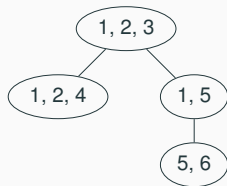
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Hypergraph



Possible hypertree decomposition

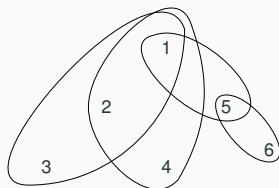


Join Trees: Hypertree Decompositions with One Bag per Hyperedge

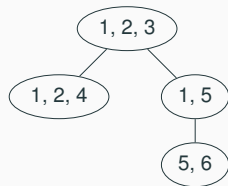
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Hypergraph



Possible hypertree decomposition



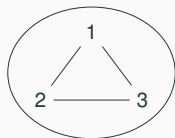
Question: Are α -acyclic precisely those hypergraphs without cycles?

α -acyclic Hypergraphs Can Have Cycles *IF* Covered by Hyperedges

Consider the FAQ:

$$\Phi() = \bigoplus_{x_1, x_2, x_3} \psi_{123}(x_1, x_2, x_3) \otimes \psi_{12}(x_1, x_2) \otimes \psi_{13}(x_1, x_3) \otimes \psi_{23}(x_2, x_3)$$

Hypergraph



Hypertree decompositions

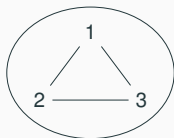


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Hypergraph



Hypertree decompositions



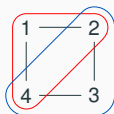
- Cycle formed by the factors ψ_{12} , ψ_{13} , and ψ_{23}
- BUT covered by the factor ψ_{123}
- We can evaluate Φ efficiently by absorbing each other factor into ψ_{123}
 $\psi_{123}(x_1, x_2, x_3) := \psi_{123}(x_1, x_2, x_3) \otimes \psi_{ij}(x_i, x_j), (i, j) \in \{(1, 2), (1, 3), (2, 3)\}$

Non-trivial α -Acyclicity Example

$$\Phi() = \bigoplus_{x_1, x_2, x_3, x_4} \psi_{124}(x_1, x_2, x_4) \otimes \psi_{234}(x_2, x_3, x_4) \otimes \psi_{12}(x_1, x_2) \otimes \psi_{23}(x_2, x_3) \otimes \psi_{34}(x_3, x_4) \otimes \psi_{14}(x_1, x_4)$$

- Cycle formed by the factors ψ_{12} , ψ_{23} , ψ_{34} , and ψ_{14}
- BUT covered by the factors ψ_{124} and ψ_{234}
- We can evaluate Φ efficiently:
 - Absorb the factors ψ_{12} and ψ_{14} into the factor ψ_{124}
 - Absorb the factors ψ_{23} and ψ_{34} into the factor ψ_{234}
 - Multiply the factors ψ_{124} and ψ_{234} and aggregate away the variables

Hypergraph

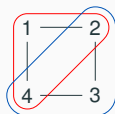


Non-trivial α -Acyclicity Example

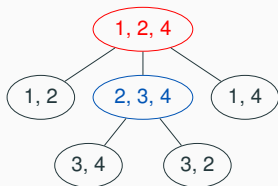
$$\Phi() = \bigoplus_{x_1, x_2, x_3, x_4} \psi_{124}(x_1, x_2, x_4) \otimes \psi_{234}(x_2, x_3, x_4) \otimes \psi_{12}(x_1, x_2) \otimes \psi_{23}(x_2, x_3) \otimes \psi_{34}(x_3, x_4) \otimes \psi_{14}(x_1, x_4)$$

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Hypergraph

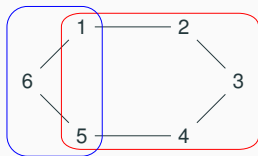


Hypertree decompositions (Second is join tree)



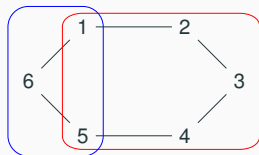
Further α -acyclicity Examples

Hypergraph

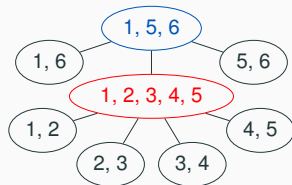


Further α -acyclicity Examples

Hypergraph

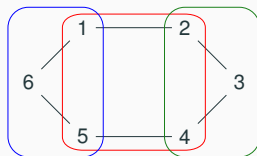
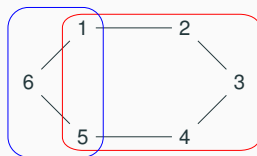


Hypertree decompositions (Join trees)

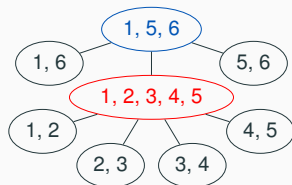


Further α -acyclicity Examples

Hypergraph

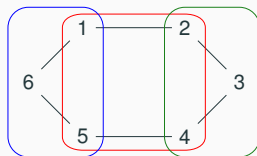
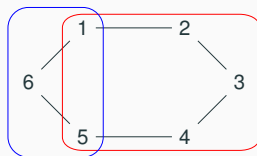


Hypertree decompositions (Join trees)

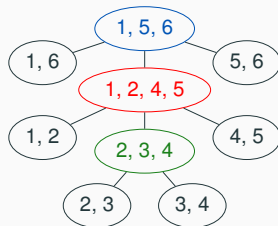
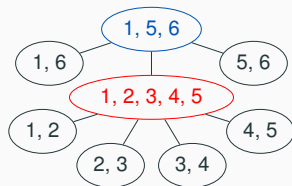


Further α -acyclicity Examples

Hypergraph



Hypertree decompositions (Join trees)



The GYO Algorithm

The GYO (Graham, Yu, Ozsoyoglu) algorithm is used to decide α -acyclicity:

Input: Hypergraph \mathcal{H}

Output: Hypergraph obtained by repeating the following rules as long as possible:

- Eliminate a node that is contained in only one hyperedge
- Eliminate a hyperedge that is contained in another hyperedge

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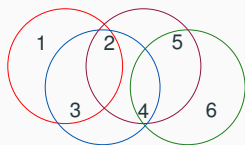
- Eliminate a node that is contained in only one hyperedge
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\mathcal{H} is α -acyclic if and only if $\text{GYO}(\mathcal{H}) = (\emptyset, \{\emptyset\})$

In words: \mathcal{H} is α -acyclic if and only if the application of GYO to \mathcal{H} returns a hypergraph with no vertices and one empty hyperedge

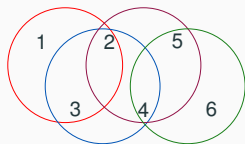
The GYO Algorithm: Example 1/3

initial hypergraph \mathcal{H}

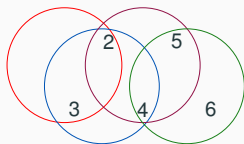


The GYO Algorithm: Example 1/3

initial hypergraph \mathcal{H}

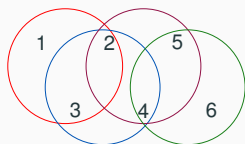


node 1 removed

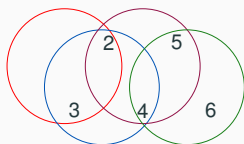


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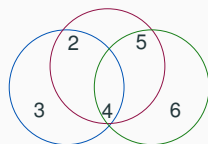
initial hypergraph \mathcal{H}



node 1 removed

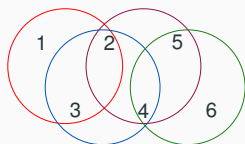


edge $\{2, 3\}$ removed

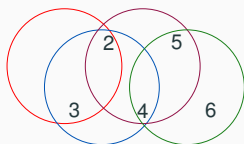


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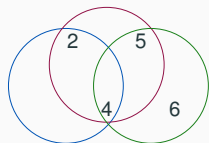
initial hypergraph \mathcal{H}



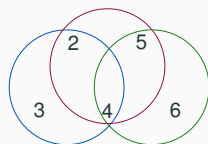
node 1 removed



node 3 removed

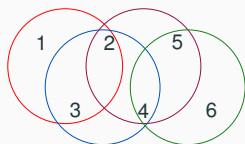


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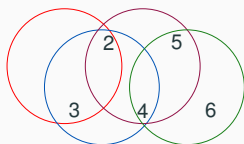


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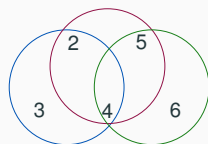
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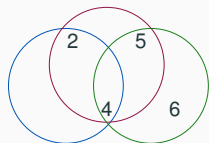
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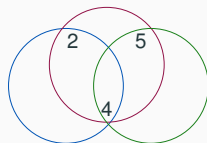
edge $\{2, 3\}$ removed



node 3 removed

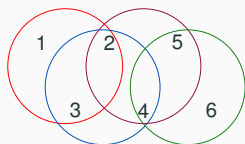


node 6 removed

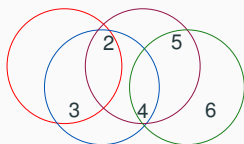


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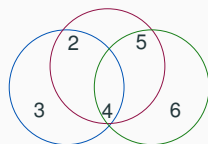
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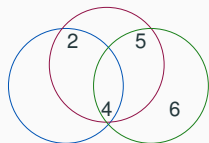
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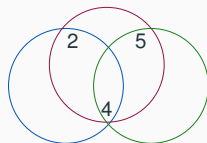
edge $\{2, 3\}$ removed



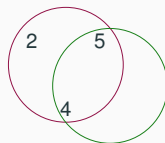
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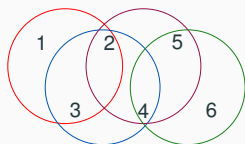


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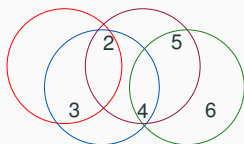


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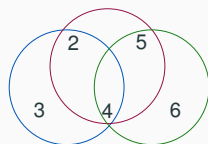
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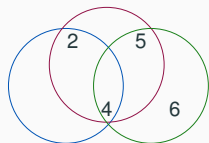
node 1 removed



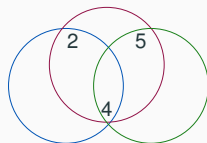
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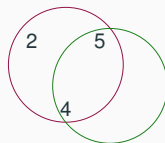
node 3 removed



node 6 removed



edge $\{2, 4\}$ removed

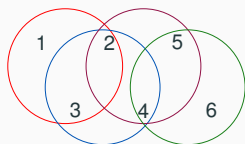


edge $\{4, 5\}$ removed

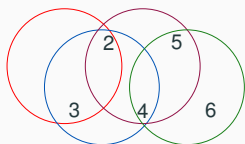


The GYO Algorithm: Example 1/3

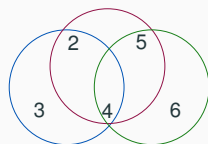
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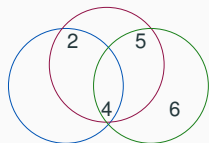
node 1 removed



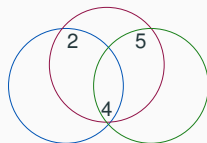
edge $\{2, 3\}$ removed



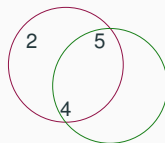
node 3 removed



node 6 removed



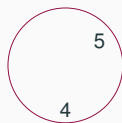
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edge $\{4, 5\}$ removed

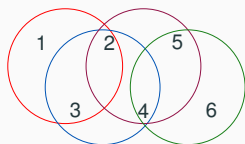


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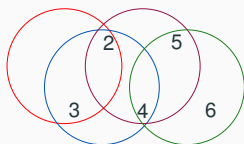


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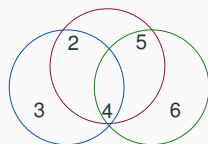
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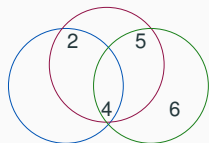
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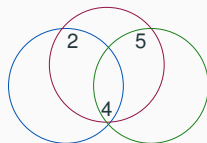
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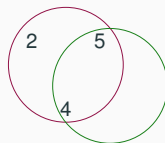
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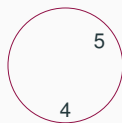
edge $\{2, 4\}$ removed



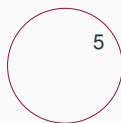
edge $\{4, 5\}$ removed



node 2 removed

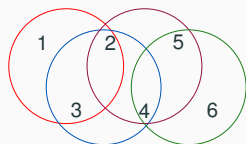


node 4 removed

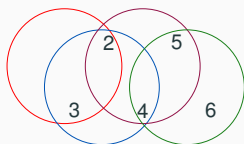


The GYO Algorithm: Example 1/3

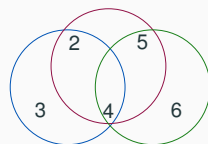
initial hypergraph \mathcal{H}



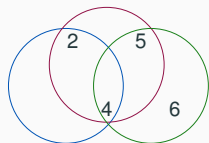
node 1 removed



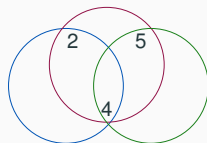
edge {2,3} removed



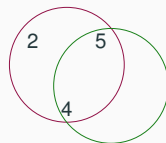
node 3 removed



node 6 removed



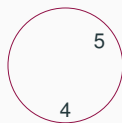
edge {2,4} removed



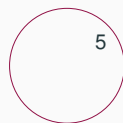
edge {4,5} removed



node 2 removed



node 4 removed



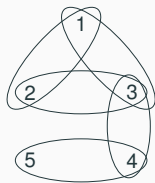
node 5 removed



$\implies \text{GYO}(\mathcal{H}) = (\emptyset, \{\emptyset\}) \implies \mathcal{H}$ is α -acyclic

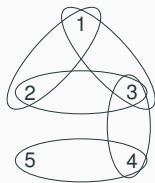
The GYO Algorithm: Example 2/3

initial hypergraph \mathcal{H}

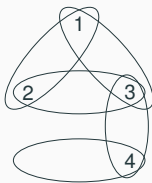


The GYO Algorithm: Example 2/3

initial hypergraph \mathcal{H}

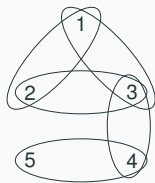


node 5 removed

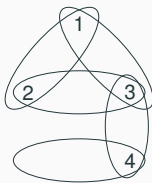


The GYO Algorithm: Example 2/3

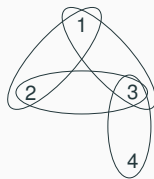
initial hypergraph \mathcal{H}



node 5 removed

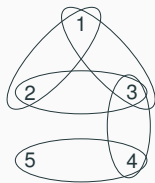


edge {4} removed

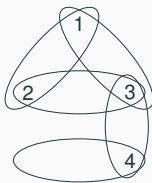


The GYO Algorithm: Example 2/3

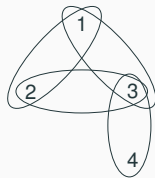
initial hypergraph \mathcal{H}



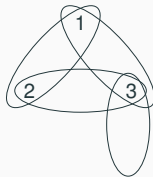
node 5 removed



edge {4} removed

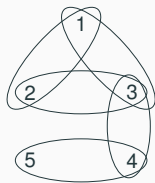


node 4 removed

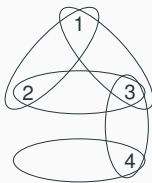


The GYO Algorithm: Example 2/3

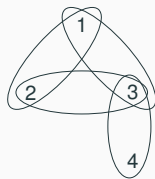
initial hypergraph \mathcal{H}



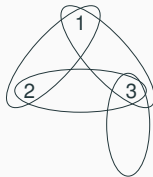
node 5 removed



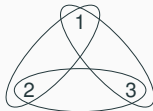
edge {4} removed



node 4 removed

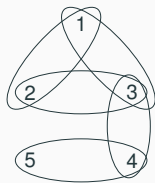


edge {3} removed

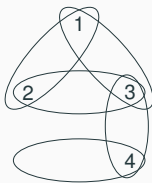


The GYO Algorithm: Example 2/3

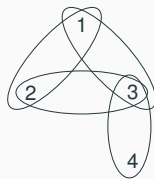
initial hypergraph \mathcal{H}



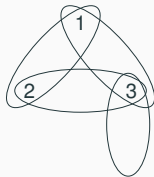
node 5 removed



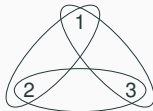
edge $\{4\}$ removed



node 4 removed



edge $\{3\}$ removed

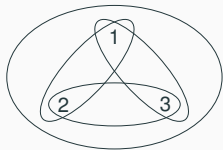


no more rule applicable

$\implies \text{GYO}(\mathcal{H}) \neq (\emptyset, \{\emptyset\}) \implies \mathcal{H}$ is not α -acyclic

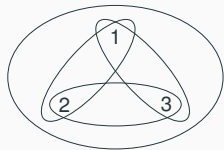
The GYO Algorithm: Example 3/3

initial hypergraph \mathcal{H}

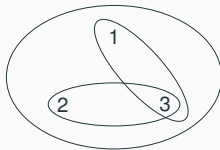


The GYO Algorithm: Example 3/3

initial hypergraph \mathcal{H}

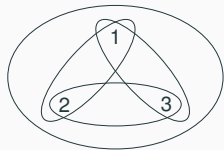


edge $\{1, 2\}$ removed

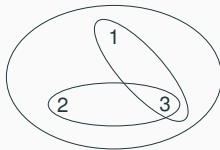


The GYO Algorithm: Example 3/3

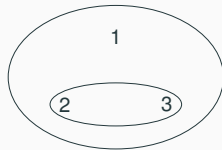
initial hypergraph \mathcal{H}



edge $\{1, 2\}$ removed

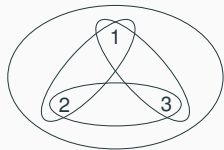


edge $\{1, 3\}$ removed

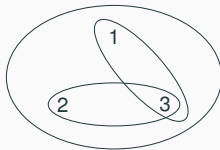


The GYO Algorithm: Example 3/3

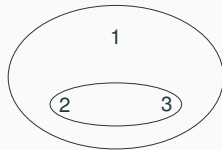
initial hypergraph \mathcal{H}



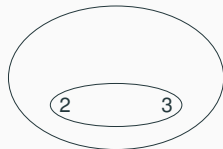
edge $\{1, 2\}$ removed



edge $\{1, 3\}$ removed

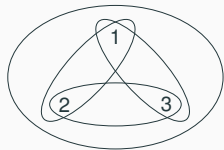


node 1 removed

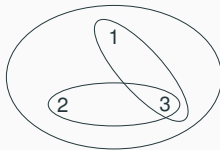


The GYO Algorithm: Example 3/3

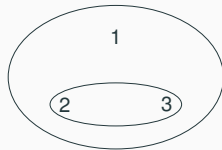
initial hypergraph \mathcal{H}



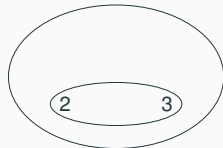
edge $\{1, 2\}$ removed



edge $\{1, 3\}$ removed



node 1 removed

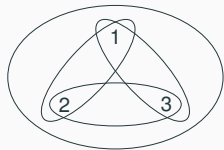


edge $\{2, 3\}$ removed

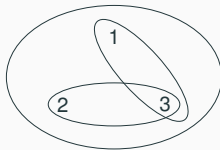


The GYO Algorithm: Example 3/3

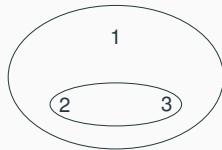
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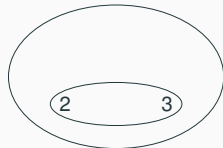
edge $\{1, 2\}$ removed



edge $\{1, 3\}$ removed



node 1 removed



edge $\{2, 3\}$ removed

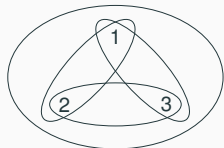


node 2 removed

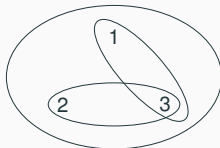


The GYO Algorithm: Example 3/3

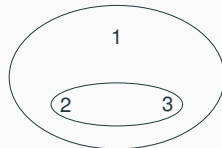
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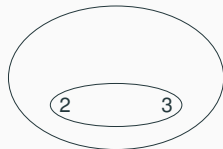
edge $\{1, 2\}$ removed



edge $\{1, 3\}$ removed



node 1 removed



edge $\{2, 3\}$ removed



node 2 removed



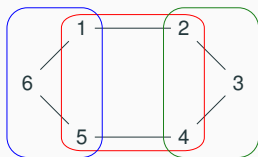
node 3 removed



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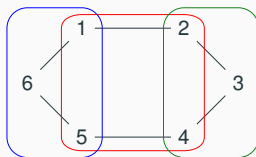
Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$



Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$

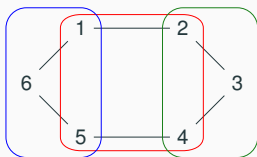


Algorithm

1. Compute *weighted graph* for \mathcal{H}

Computing a Join Tree for an α -Acyclic Hypergraph

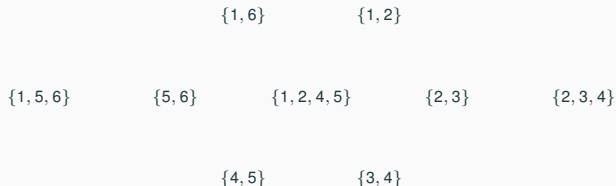
Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$



Algorithm

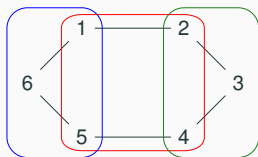
1. Compute *weighted graph* for \mathcal{H}
 - vertex set: \mathcal{E}

Weighted graph for \mathcal{H} :



Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$

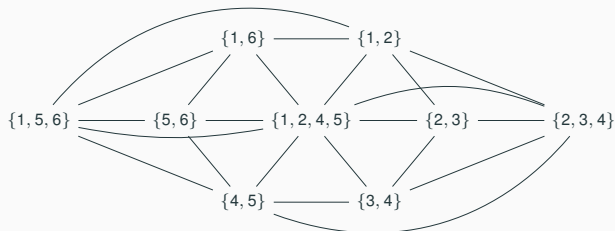


Algorithm

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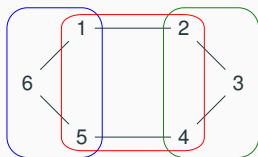
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Weighted graph for \mathcal{H} :



Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$

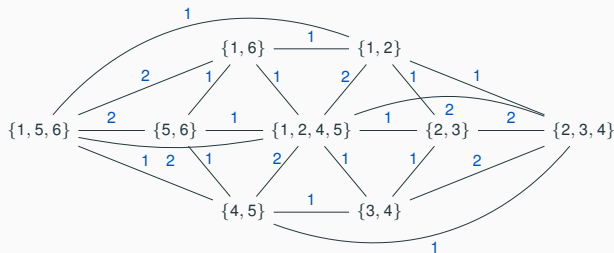


Algorithm

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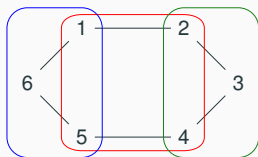
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Weighted graph for \mathcal{H} :



Computing a Join Tree for an α -Acyclic Hypergraph

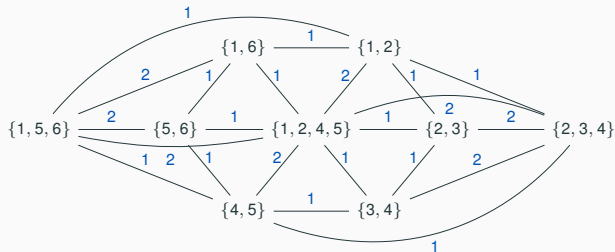
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Algorithm

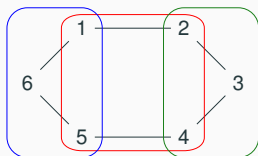
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2. Compute a maximum-weight spanning tree, e.g. using Prim's Algorithm

Weighted graph for \mathcal{H} :



Computing a Join Tree for an α -Acyclic Hypergraph

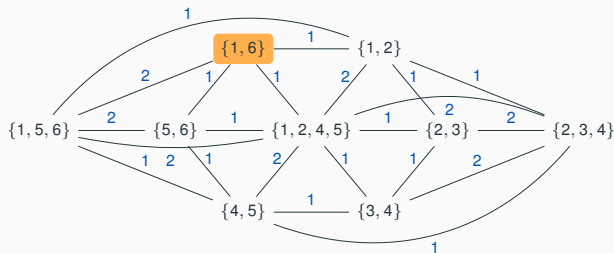
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Algorithm

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 - Start with $S = \{e\}$ for an arbitrary $e \in \mathcal{E}$

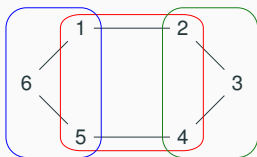
Weighted graph for \mathcal{H} :



S

Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$



Algorithm

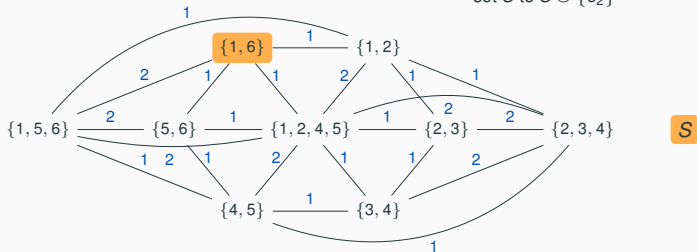
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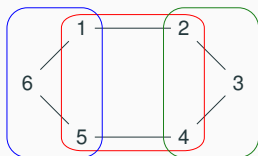
- Start with $S = \{e\}$ for an arbitrary $e \in \mathcal{E}$
- While $S \neq \mathcal{E}$:
 - choose edge (e_1, e_2) with $e_1 \in S, e_2 \notin S$ with maximal weight
 - (e_1, e_2) becomes spanning tree edge
 - set S to $S \cup \{e_2\}$

Weighted graph for \mathcal{H} :



Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$



Algorithm

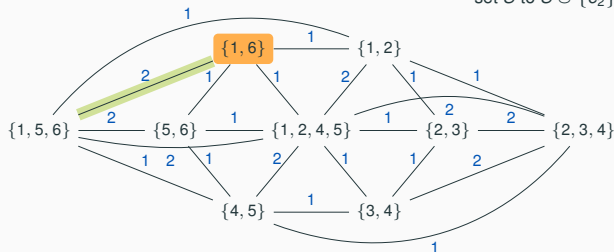
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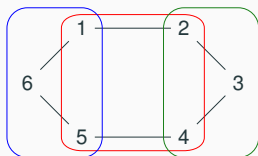
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Weighted graph for \mathcal{H} :



Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$



Algorithm

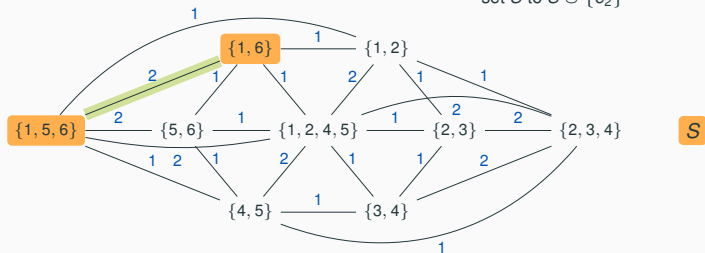
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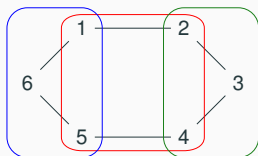
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Weighted graph for \mathcal{H} :



Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$



Algorithm

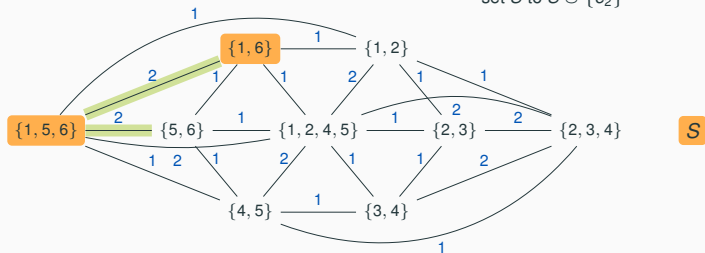
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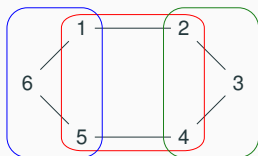
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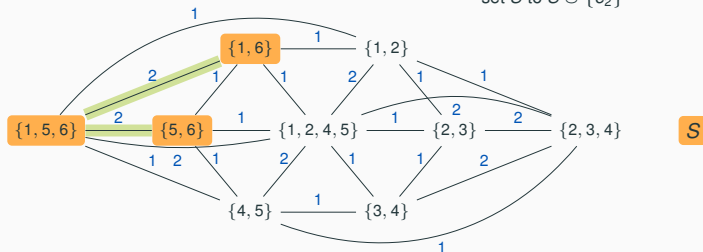
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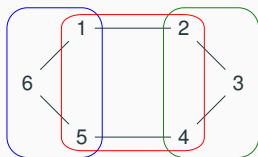
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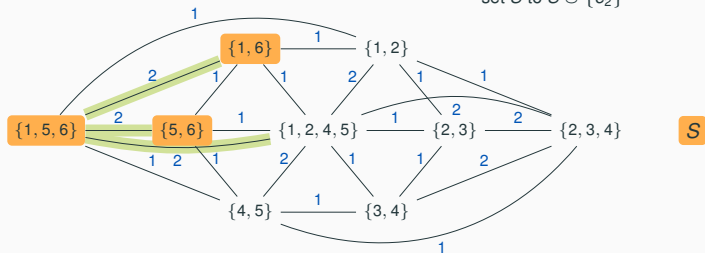
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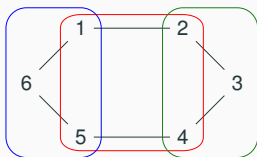
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Computing a Join Tree for an α -Acyclic Hypergraph

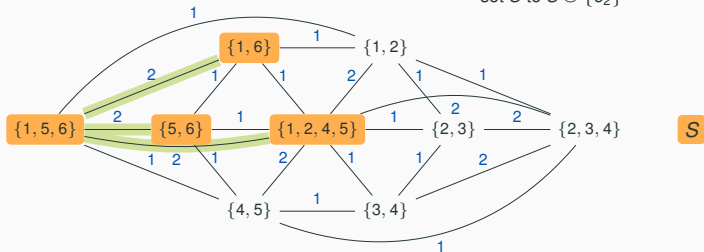
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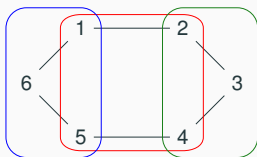
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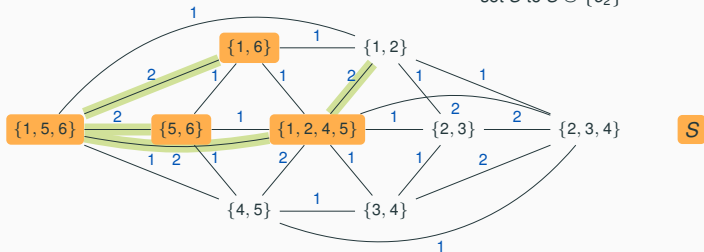
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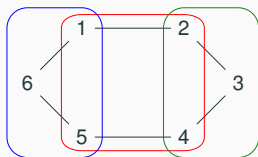
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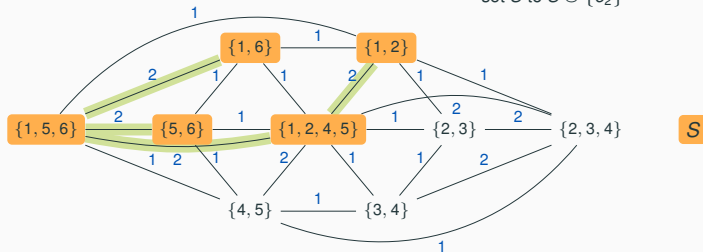
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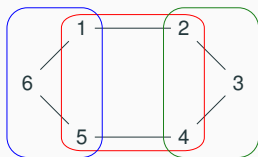
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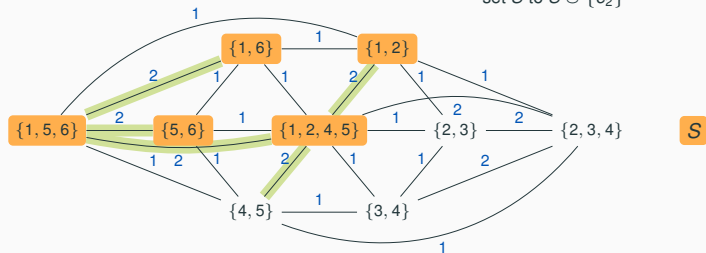
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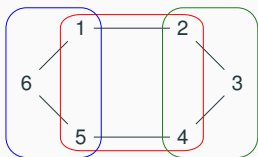
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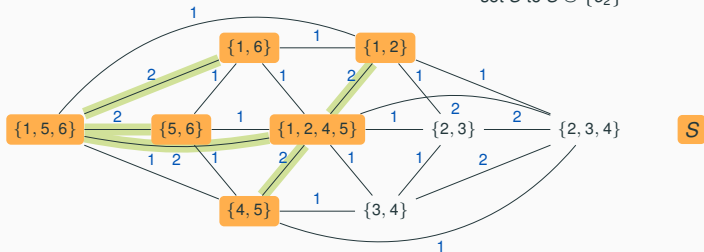
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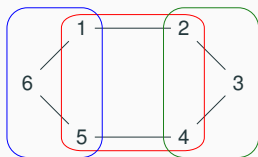
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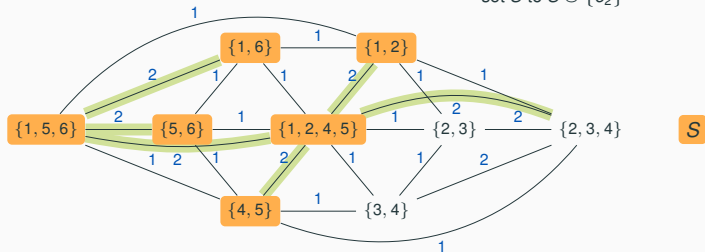
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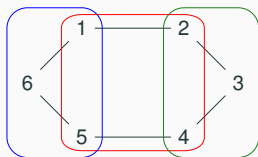
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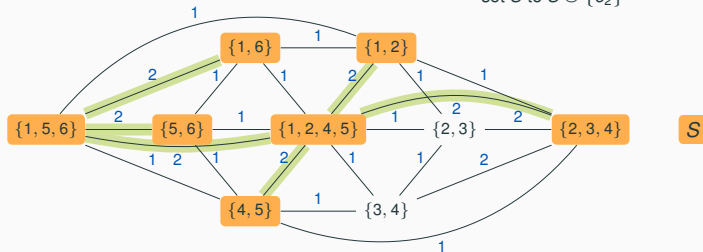
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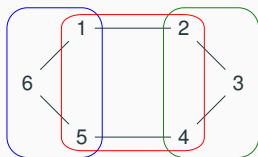
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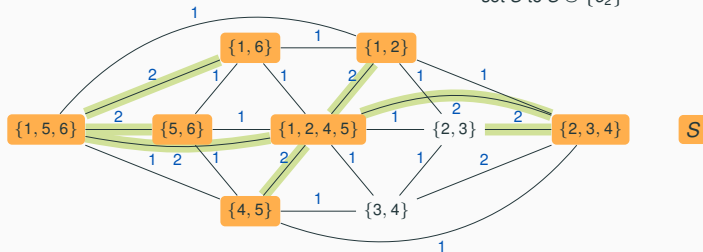
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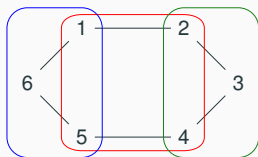
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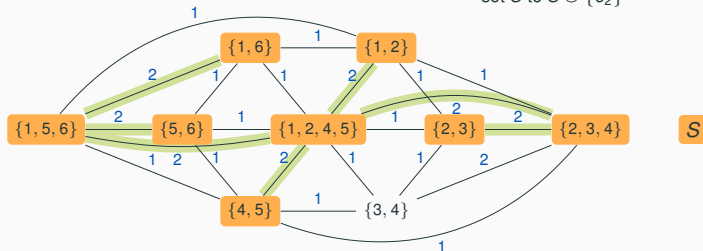
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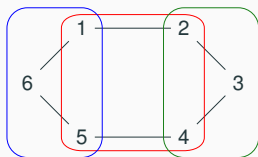
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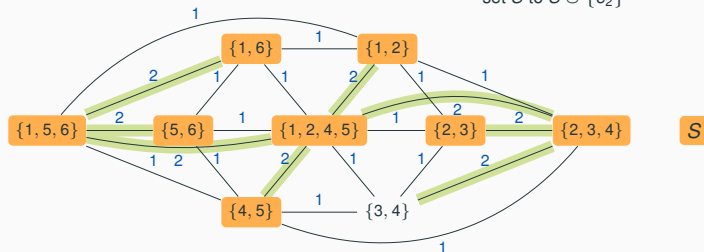
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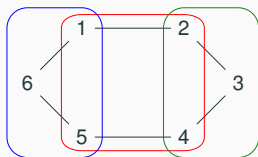
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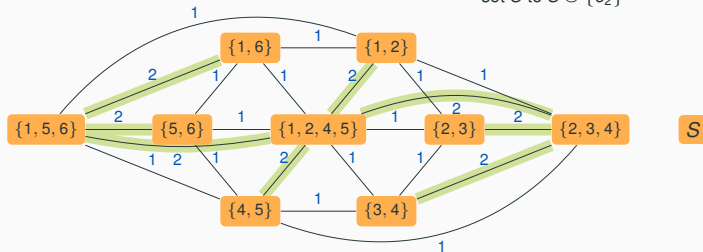
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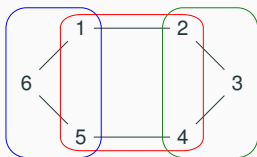
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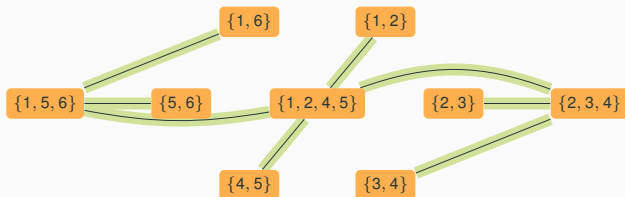
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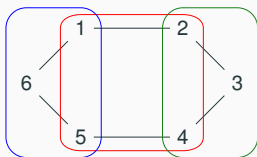
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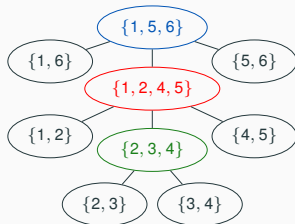


Computing a Join Tree for an α -Acyclic Hypergraph

Hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$



Join tree for \mathcal{H} :



Algorithm

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β -Acyclicity: Every Subgraph is Acyclic

β -acyclicity is closed under hyperedge removal, α -acyclicity is not closed

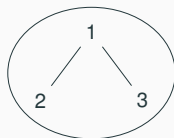
- α -acyclic hypergraphs may have cycles covered by hyperedges
 - The hypergraph without such hyperedges is not α -acyclic
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 - Removal of any hyperedges preserves β -acyclicity

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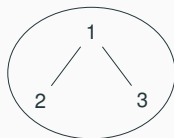


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Hypergraph



Hypertree decompositions



Free-connex α -Acyclicity

- α -acyclicity is prerequisite to efficient computation
- FAQ compute time also depends on free variables X_1, \dots, X_f

$$\Phi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1}}^{(f+1)} \cdots \bigoplus_{x_n}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

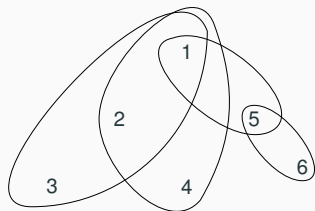
- **Free-connex property:** There is a join tree for the hypergraph of Φ , where
 - All free variables $[f]$ appear in nodes that form a connected subtree
 - Each bound variable either occurs in nodes without free variables or only in one node with free variables
 - \rightarrow All bound variables can be aggregated away in linear time

When is an FAQ with hypergraph \mathcal{H} and free variables $[f]$ free-connex α -acyclic?

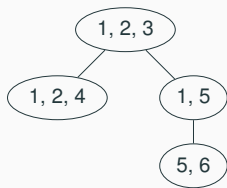
\mathcal{H} remains α -acyclic even after adding the hyperedge $[f]$ over the free variables

Free-connex α -Acyclicity Example 1/2

α -acyclic hypergraph \mathcal{H}

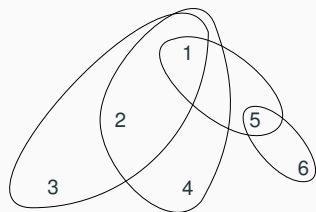


Possible join tree for \mathcal{H}

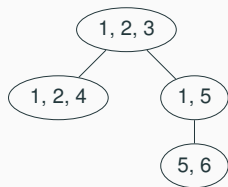


Free-connex α -Acyclicity Example 1/2

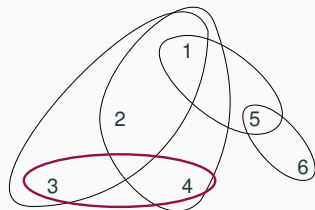
α -acyclic hypergraph \mathcal{H}



Possible join tree for \mathcal{H}

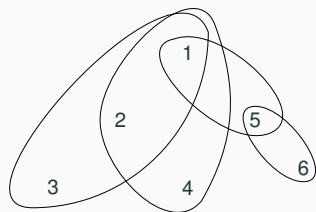


choose $\{3, 4\}$
as free variables

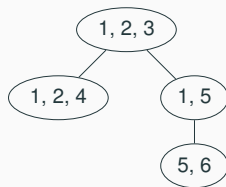


Free-connex α -Acyclicity Example 1/2

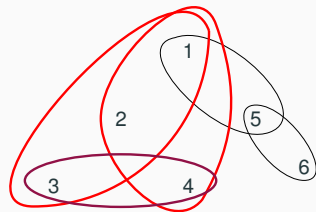
α -acyclic hypergraph \mathcal{H}



Possible join tree for \mathcal{H}



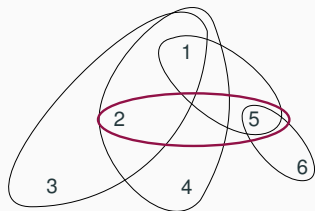
choose $\{3, 4\}$
as free variables



- \Rightarrow Add the hyperedge $\{3, 4\}$ to the hypergraph
- \Rightarrow There is a cycle not covered by a hyperedge
- \Rightarrow Extended hypergraph not α -acyclic
- \Rightarrow No join tree for extended hypergraph
- \Rightarrow Original hypergraph not free-connex α -acyclic

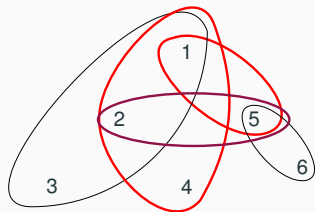
Free-connex α -Acyclicity Example 2/2

choose $\{2, 5\}$
as free variables



Free-connex α -Acyclicity Example 2/2

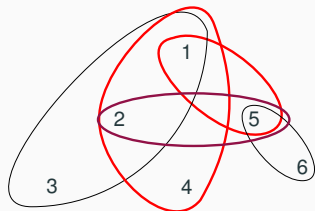
choose $\{2, 5\}$
as free variables



- \Rightarrow Add the hyperedge $\{2, 5\}$ to the hypergraph
- \Rightarrow There is a cycle not covered by a hyperedge
- \Rightarrow Extended hypergraph not α -acyclic
- \Rightarrow No join tree for extended hypergraph
- \Rightarrow Original hypergraph not free-connex α -acyclic

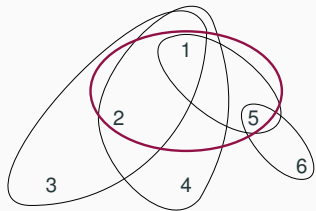
Free-connex α -Acyclicity Example 2/2

choose $\{2, 5\}$
as free variables



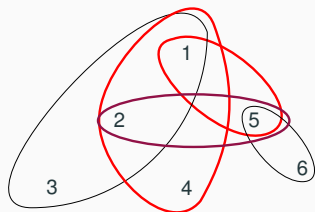
- \Rightarrow Add the hyperedge $\{2, 5\}$ to the hypergraph
- \Rightarrow There is a cycle not covered by a hyperedge
- \Rightarrow Extended hypergraph not α -acyclic
- \Rightarrow No join tree for extended hypergraph
- \Rightarrow Original hypergraph not free-connex α -acyclic

choose $\{1, 2, 5\}$
as free variables



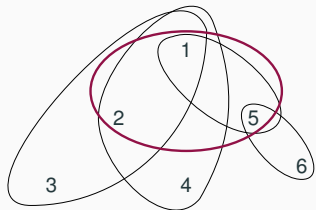
Free-connex α -Acyclicity Example 2/2

choose $\{2, 5\}$
as free variables

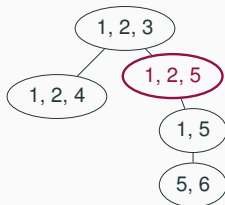


- \Rightarrow Add the hyperedge $\{2, 5\}$ to the hypergraph
- \Rightarrow There is a cycle not covered by a hyperedge
- \Rightarrow Extended hypergraph not α -acyclic
- \Rightarrow No join tree for extended hypergraph
- \Rightarrow Original hypergraph not free-connex α -acyclic

choose $\{1, 2, 5\}$
as free variables



Possible join tree for the extended hypergraph



- \Rightarrow Extended hypergraph α -acyclic
- \Rightarrow Original hypergraph
free-connex α -acyclic

Landscape of Hypergraph Types

