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Reformulating $R(*,m)C$ with Tree Decomposition

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Reformulating $R(*,m)C$ with Tree Decomposition

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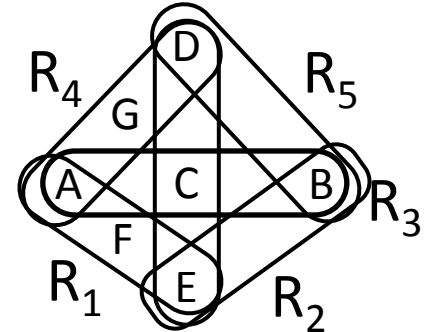
Outline

- Introduction
- $R(*,m)C$ Property & Algorithm
- Exploit Tree Decomposition to
 - Avoid useless update & reduce propagation effort
 - ↪ Update queue: $PROCESSQ \rightsquigarrow PROCESSMQ$
 - ↪ The two algorithms *yield the same filtering*
 - Synthesize & add new constraints to improve propagation
 - ↪ Property enforced: $R(*,m)C \rightsquigarrow T-R(*,m,z)C$
 - ↪ The same algorithm *yields stronger filtering*
- Experimental Results
- Conclusion

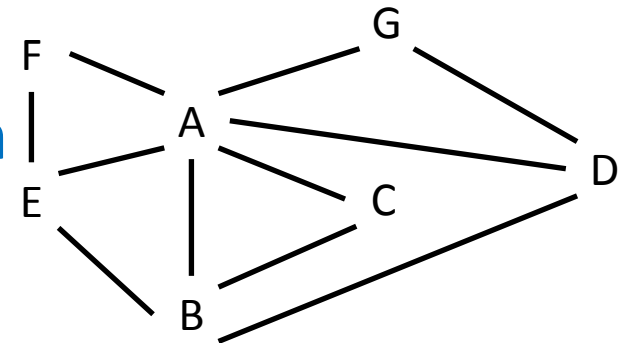
Constraint Satisfaction Problem

- CSP
 - Variables (\mathcal{V}), domains
 - Constraints: relations (\mathcal{R}), scope
- Representation
 - Hypergraph
 - Primal graph
 - Dual graph
- Solved with
 - Search
 - Enforcing consistency
- Warning
 - Consistency property vs. algorithms

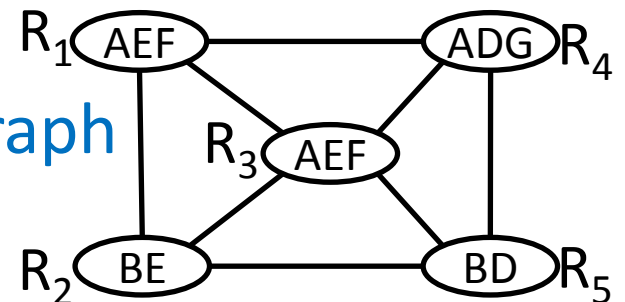
Hypergraph



Primal graph

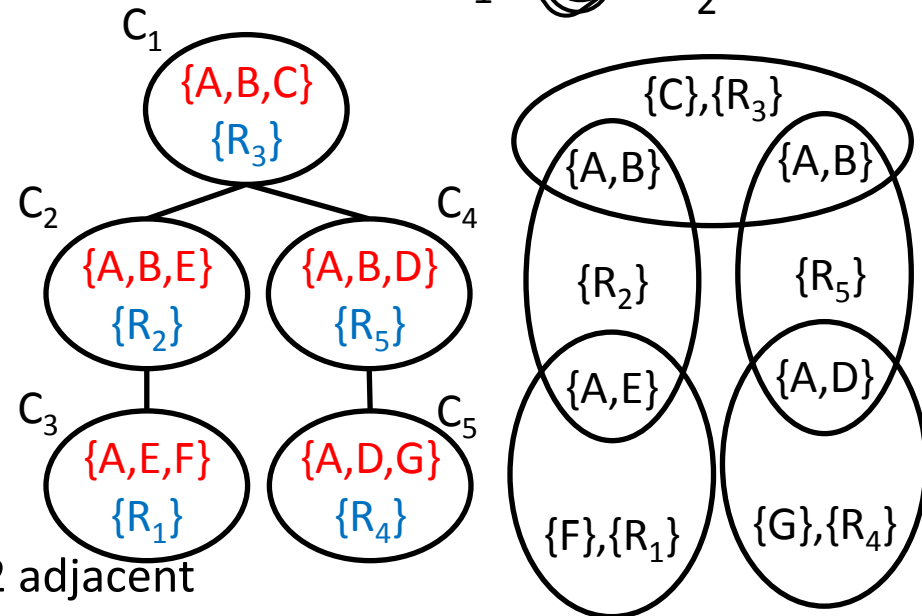
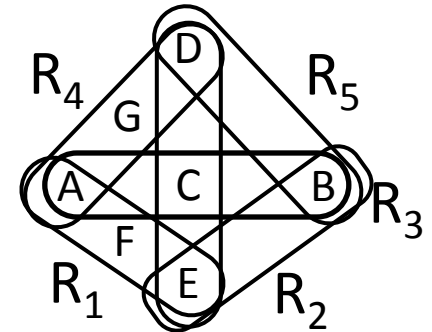


Dual graph



Tree Decomposition

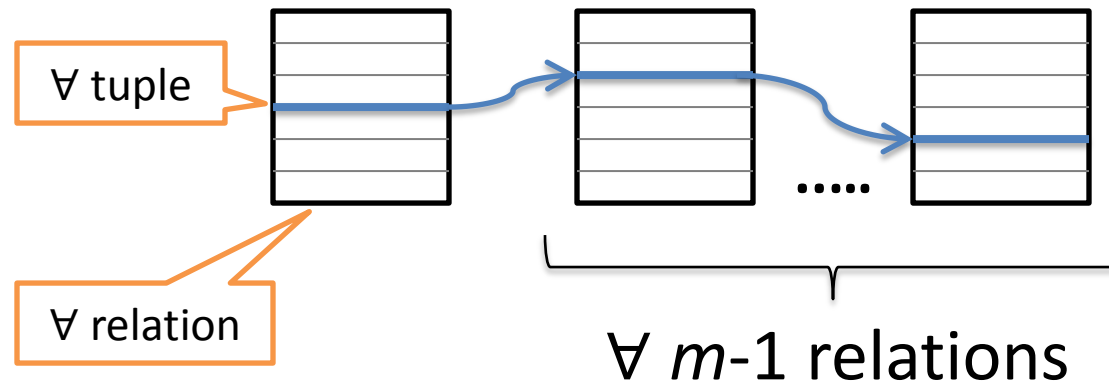
- Tree: Vertices/clusters, edges
- Each cluster is labeled with
 - A set of variables $\subseteq \mathcal{V}$
 - A set of relations $\subseteq \mathcal{R}$
- Two conditions
 1. For each relation R , \exists cluster c_i
 - R appears in c_i
 - $\text{Scope}(R)$ is also in c_i
 2. Every variable
 - Induces a connected subtree
- Separators
 - Variables & relations common to 2 adjacent clusters
 - channel communications between clusters



$R(*,m)C$ Property

[Karakashian+ 10]

- A CSP is $R(*,m)C$ iff
 - Every **tuple** in a relation can be extended to the variables in the scope of any $(m-1)$ other relations in an assignment satisfying all m relations simultaneously



ProcessQ: Algorithm for $R(*,m)C$

- Φ : combination of m connected relations in the dual graph

$$\Phi = \{ \omega_1 = \{R_1, R_2, \dots, R_m\}, \omega_2, \omega_3, \dots, \omega_k \}$$

- Q propagation queue

$$Q = \{ \langle R_1, \omega_1 \rangle, \langle R_1, \omega_2 \rangle, \langle R_1, \omega_3 \rangle, \dots, \langle R_n, \omega_{k-1} \rangle, \langle R_n, \omega_k \rangle \}$$

- For each $\langle R_i, \omega_j \rangle$ in Q, ProcessQ
 - Deletes from R_i tuples that cannot be extended to relations in ω_j
 - As some tuples of relations $R_x \in \omega_j$ may lose support, it requeues $\{ \langle R_x, \omega_y \rangle \}$ for every threatened relation

ProcessQ: Animation

Q
$\langle R_1, \omega_1 \rangle$
$\langle R_2, \omega_1 \rangle$
$\langle R_5, \omega_1 \rangle$
$\langle R_2, \omega_2 \rangle$
$\langle R_5, \omega_2 \rangle$
$\langle R_4, \omega_2 \rangle$
$\langle R_3, \omega_3 \rangle$
$\langle R_4, \omega_3 \rangle$
$\langle R_5, \omega_3 \rangle$

Extract $\langle R, \omega \rangle$ from Q

Define CSP P_ω

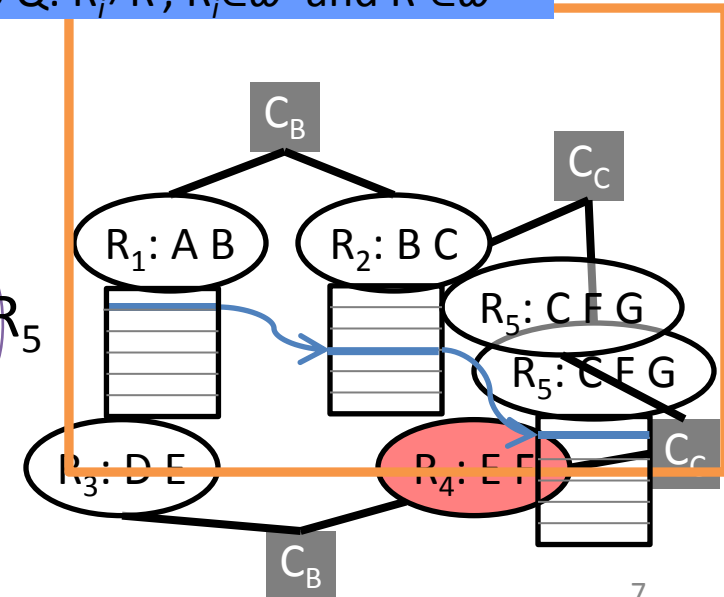
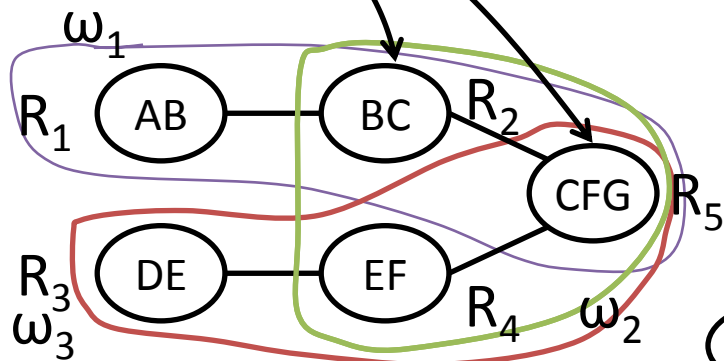
For each τ in R

Assign τ as a value for R

Solve P_ω with forward checking

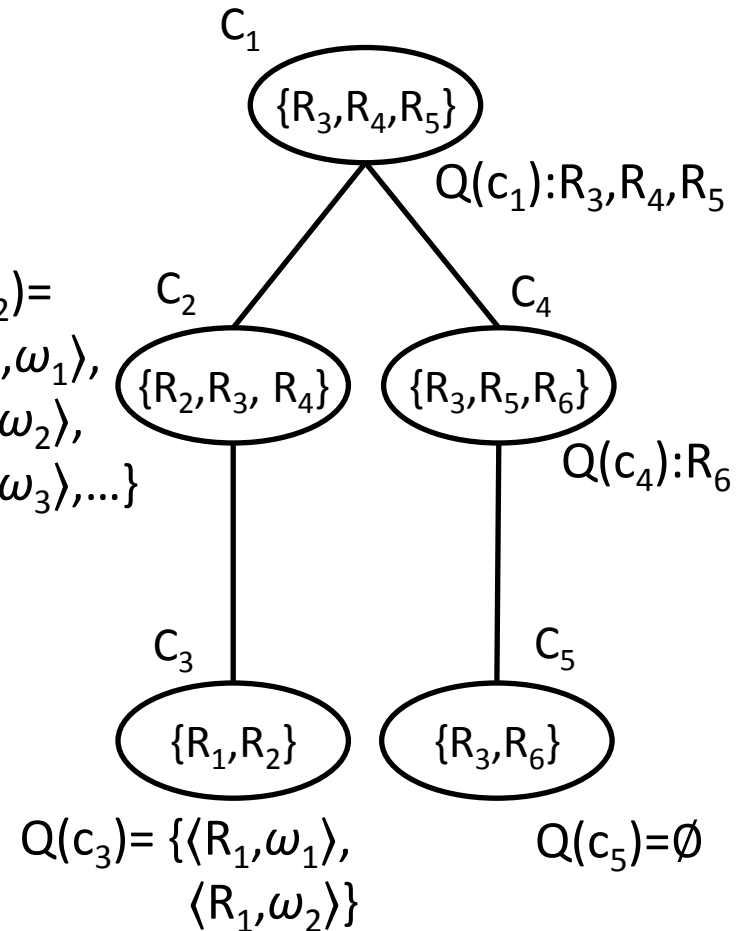
If no solution found: delete τ

Add $\langle R', \omega' \rangle$ to Q: $R_i \neq R', R_i \in \omega'$ and $R' \in \omega'$



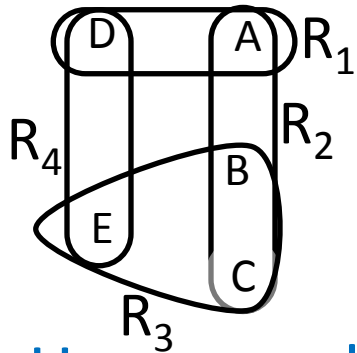
ProcessMQ: Intelligent update scheduling

- Cluster c_i has a local queue $Q(c_i)=\{\langle R_i, \omega \rangle\}$ for relations R_i in cluster but not in parent
- Using the tree decomposition
 - As an ordering heuristic for checking consistency of $\langle R_i, \omega \rangle$
 - Repeat “leaves up to root, down to leaves,” until quiescence
 - Update relations in only local queue
 - Example: R_3 is updated only when root is reached
- Advantage fewer updates, same filtering
 - In previous example, R_3 is updated once although it appears in 3 clusters

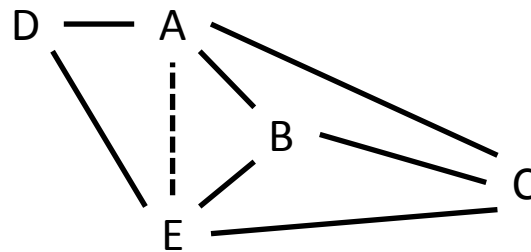


T-R(*,m,z)C

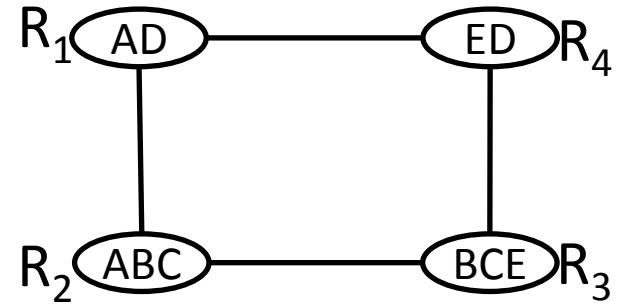
[Rollon+ 10]



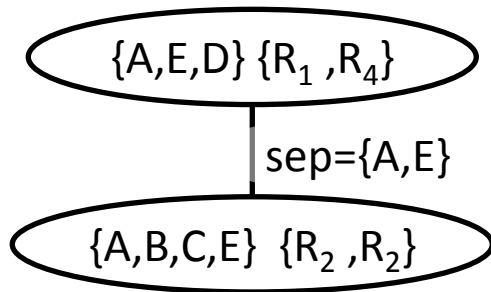
Hypergraph



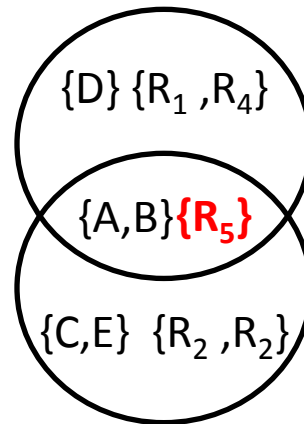
Primal graph



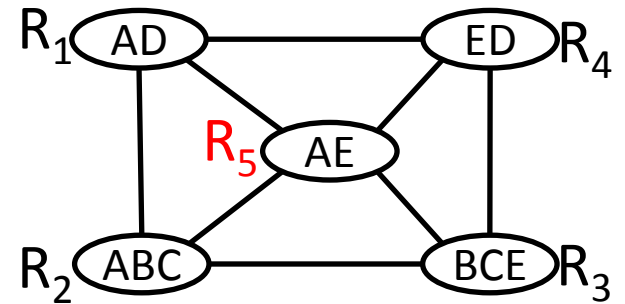
Dual graph



Tree decomposition



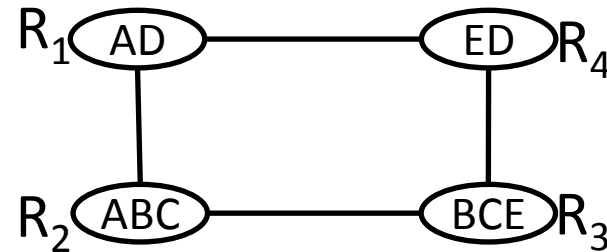
Adding R_5



T-R(*,m,z)C Strictly Stronger than R(*,m)C

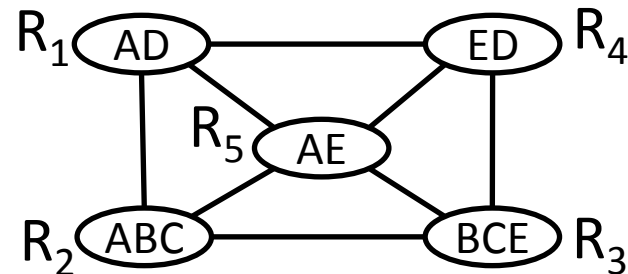
Let A, B, C, D and E be Boolean variables

R_1	R_2	R_3	R_4
A D	A B C	B C E	E D
0 0	0 0 0	0 0 0	0 0
1 1	1 1 1	1 1 1	1 1



Assignment A=0 & E=1 is valid
Does not violate R(*,2)C

R_5
A E
0 0
1 1



Assignment A=0 & E = 1 is **inconsistent**

Experimental Results

- Experiments for finding all solutions with BTD maintaining $wR(*, \text{best}(2,3,4))C$ and $T\text{-}wR(*, \text{best}(2,3,4), \text{best}(5,7,9))$
- Results shown demonstrate the benefits of ProcessMQ & $T\text{-}wR(*, m, z)C$

Benchmark	#ins	#vars	tw		ProcessQ $wR(*, \text{best})C$	ProcessMQ $wR(*, \text{best})C$	ProcessQ $T\text{-}wR(*, b, b)C$
aim-200	24	200	104.92	#C	17	17	<u>22</u>
				t_{avg}	246.35	252.48	<u>238.99</u>
				t_{max}	3,352.54	3,452.98	<u>1,540.94</u>
ogdVg	59	134	85	#C	15	15	15
				t_{avg}	283.27	<u>242.06</u>	266.74
				t_{max}	1,834.11	<u>1,508.27</u>	1,720.97
rand-3-20-20	50	20	13	#C	13	<u>14</u>	-
				t_{avg}	2,191.56	<u>1,949.87</u>	-
				t_{max}	3,481.04	<u>3,145.77</u>	-

Conclusions

- Contributions
 - Reformulated $R(*,m)C$ algorithm
 - New relational consistency property $T-R(*,m,z)C$
 - Experimental analysis
- Future work
 - Study impact of choice of parameters z, m
 - Develop strategies for dynamically choosing z, m as a function of the size of clusters & separators