Foundations of Artificial Intelligence D7. Constraint Satisfaction Problems: Decomposition Methods

Malte Helmert

University of Basel

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M. Helmert (University of Basel)

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Foundations of Artificial Intelligence April 17, 2024 — D7. Constraint Satisfaction Problems: Decomposition Methods

D7.1 Decomposition Methods

D7.2 Conditioning

D7.3 Tree Decomposition

D7.4 Summary

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Constraint Satisfaction Problems: Overview

Chapter overview: constraint satisfaction problems

- D1–D2. Introduction
- ▶ D3–D5. Basic Algorithms
- D6–D7. Problem Structure
 - D6. Constraint Graphs
 - ▶ D7. Decomposition Methods

D7.1 Decomposition Methods

More Complex Graphs

What if the constraint graph is not a tree and does not decompose into several components?

- idea 1: conditioning
- idea 2: tree decomposition

German: Konditionierung, Baumzerlegung

D7.2 Conditioning

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Conditioning

Conditioning

idea: Apply backtracking with forward checking until the constraint graph restricted to the remaining unassigned variables decomposes or is a tree.

remaining problem ~> algorithms for simple constraint graphs

cutset conditioning:

Choose variable order such that early variables form a small cutset (i.e., set of variables such that removing these variables results in an acyclic constraint graph).

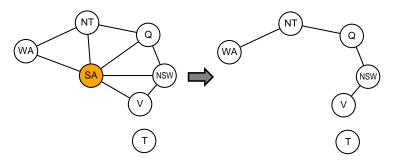
German: Cutset

time complexity: *n* variables, m < n in cutset, maximal domain size *k*: $O(k^m \cdot (n-m)k^2)$

(Finding optimal cutsets is an NP-complete problem.)

Conditioning: Example

Australia example: Cutset of size 1 suffices:



D7.3 Tree Decomposition

Tree Decomposition

basic idea of tree decomposition:

- Decompose constraint network into smaller subproblems (overlapping).
- Find solutions for the subproblems.
- Build overall solution based on the subsolutions.

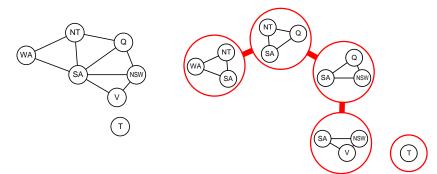
more details:

- "Overall solution building problem" based on subsolutions is a constraint network itself (meta constraint network).
- Choose subproblems in a way that the constraint graph of the meta constraint network is a tree/forest.
 violation with efficient tree algorithm

Tree Decomposition: Example



tree decomposition:



Tree Decomposition: Definition

Definition (tree decomposition)

Consider a constraint network C with variables V.

A tree decomposition of $\ensuremath{\mathcal{C}}$

is a graph $\ensuremath{\mathcal{T}}$ with the following properties.

requirements on vertices:

- Every vertex of T corresponds to a subset of the variables V. Such a vertex (and corresponding variable set) is called a subproblem of C.
- Every variable of V appears in at least one subproblem of T.
- For every nontrivial constraint R_{uv} of C, the variables u and v appear together in at least one subproblem in T.

. . .

Tree Decomposition: Definition

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Definition (tree decomposition)
Consider a constraint network C with variables V.
A tree decomposition of C
is a graph \mathcal{T} with the following properties.
. . .
requirements on edges:
  For each variable v \in V, let \mathcal{T}_{v} be the set of vertices
     corresponding to the subproblems that contain v.
  For each variable v, the set \mathcal{T}_{v} is connected,
     i.e., each vertex in \mathcal{T}_{v} is reachable from every other vertex
     in \mathcal{T}_{\nu} without visiting vertices not contained in \mathcal{T}_{\nu}.

    T is acyclic (a tree/forest)
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Meta Constraint Network

meta constraint network $C^{T} = \langle V^{T}, dom^{T}, (R^{T}_{uv}) \rangle$ based on tree decomposition T

- $V^{\mathcal{T}}$:= vertices of \mathcal{T} (i.e., subproblems of \mathcal{C} occurring in \mathcal{T})
- dom^T(v) := set of solutions of subproblem v
- R^T_{uv} := {⟨s, t⟩ | s, t compatible solutions of subproblems u, v}
 if {u, v} is an edge of T. (All constraints between
 subproblems not connected by an edge of T are trivial.)

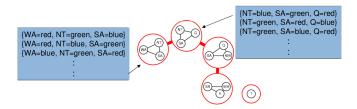
German: Meta-Constraintnetz

Solutions of two subproblems are called compatible if all overlapping variables are assigned identically.

Solving with Tree Decompositions: Algorithm

algorithm:

- Find all solutions for all subproblems in the decomposition and build a tree-like meta constraint network.
- Constraints in meta constraint network: subsolutions must be compatible.
- Solve meta constraint network with an algorithm for tree-like networks.



Good Tree Decompositions

goal: each subproblem has as few variables as possible

- crucial: subproblem V' in \mathcal{T} with highest number of variables
- number of variables in V' minus 1 is called width of the decomposition
- best width over all decompositions: tree width of the constraint graph (computation is NP-complete)

time complexity of solving algorithm based on tree decompositions: $O(nk^{w+1})$, where *w* is width of decomposition (requires specialized version of revise; otherwise $O(nk^{2w+2})$.)

D7.4 Summary

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Summary: This Chapter

- Reduce complex constraint graphs to simple constraint graphs.
- cutset conditioning:
 - Choose as few variables as possible (cutset) such that an assignment to these variables yields a remaining problem which is structurally simple.
 - search over assignments of variables in cutset

tree decomposition: build tree-like meta constraint network

- meta variables: groups of original variables that jointly cover all variables and constraints
- values correspond to consistent assignments to the groups
- constraints between overlapping groups to ensure compatibility
- overall algorithm exponential in width of decomposition (size of largest group)

Summary: CSPs

Constraint Satisfaction Problems (CSP)

General formalism for problems where

- values have to be assigned to variables
- such that the given constraints are satisfied.
- algorithms: backtracking search + inference (e.g., forward checking, arc consistency, path consistency)
- variable and value orders important
- more efficient: exploit structure of constraint graph (connected components; trees)

More Advanced Topics

more advanced topics (not considered in this course):

- backjumping: backtracking over several layers
- no-good learning: infer additional constraints based on information collected during backtracking
- local search methods in the space of total, but not necessarily consistent assignments
- tractable constraint classes: identification of constraint types that allow for polynomial algorithms
- solutions of different quality: constraint optimization problems (COP)

 \rightsquigarrow more than enough content for a one-semester course