

Constraint Satisfaction Problems

Constraint Networks

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October 27, 2014

1 Constraint Networks



- Constraint networks
- Solution
- Normalized Constraint Networks
- Deduction

Constraint Networks

Constraint networks
Solution
Normalized Constraint Networks
Deduction

Constraint Networks and Graphs

Solving Constraint Networks

October 27, 2014

Wöfl, Nebel and Becker-Asano – Constraint Satisfaction Problems

3 / 27

Constraint networks



Definition

A **constraint network** is a triple

$$N = \langle V, \text{dom}, C \rangle$$

where:

- V is a non-empty and finite set of **variables**;
- dom is a function that assigns to each variable $v \in V$ a non-empty set $\text{dom}(v)$ ($\text{dom}(v)$ is called the **domain** of v , elements of $\text{dom}(v)$ are called **values**);
- C is a set of relations over variables of V (called **constraints**), i.e., each constraint is a relation R_{x_1, \dots, x_m} over some scheme $S = (x_1, \dots, x_m)$ of variables in V .

The set of constraint schemes $\{S_1, \dots, S_t\}$ is called **network scheme**.

Constraint Networks
Constraint networks
Solution
Normalized Constraint Networks
Deduction
Constraint Networks and Graphs
Solving Constraint Networks

October 27, 2014

Wöfl, Nebel and Becker-Asano – Constraint Satisfaction Problems

4 / 27

Constraint networks



If we assume some ordering of the variables in V , we can write networks more compactly:

Definition

A **constraint network** is a triple $N = \langle V, D, C \rangle$ where:

- $V = \{v_1, \dots, v_n\}$ is a non-empty and finite ordered set of **variables** (assume order (v_1, \dots, v_n));
- $D = (D_1, \dots, D_n)$ is a sequence of **domains** for V (with D_i the domain of variable v_i and $D_N = D_1 \times \dots \times D_n$ is the **domain** of N);
- C is a set of **constraints** (x, R) where $x = (v_{i_1}, \dots, v_{i_m})$ is a scheme of variables in V and $R \subseteq D_{i_1} \times \dots \times D_{i_m}$.

Notice: any ordering of the variables suffices. Constraint network that differ only in the variable ordering are considered equal.

Constraint Networks
Constraint networks
Solution
Normalized Constraint Networks
Deduction
Constraint Networks and Graphs
Solving Constraint Networks

October 27, 2014

Wöfl, Nebel and Becker-Asano – Constraint Satisfaction Problems

5 / 27

Constraint networks



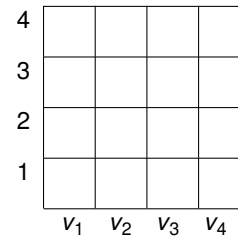
- Note that we consider **finitely** many variables only, but (e.g., for theoretical studies) this could be relaxed.
- In the 2nd definition we assumed that the relations of the constraint are embedded in the domain of the network: $R \subseteq D_{i_1} \times \dots \times D_{i_m}$. Such networks are called **embedded networks** (see Bessiere, 2006).
- The definition does not require that constraint relations are given explicitly (**in extension**, i.e. by a set of its tuples; **table constraint**). A constraint relation R could be specified by any Boolean function f_R : a tuple satisfies the constraint R iff f_R applied on the tuple gives 1.
- If not stated otherwise, we will always assume that the domains of the variables are given **in extension**.

Constraint Networks
 Constraint networks
 Solution
 Normalized Constraint Networks
 Deduction
 Constraint Networks and Graphs
 Solving Constraint Networks

Example: 4-queens problem



The 4-queens problem can be represented as a constraint network. For example, consider variables v_1, \dots, v_4 (each associated to a column of the 4×4 -chess board). Each variable v_i has as its domain $D_i = \{1, \dots, 4\}$ (conceived of as the row positions of a queen in column i).



Define then binary constraints (thus encoding "non-attacking queen positions"):
 $R_{v_1, v_2} := \{(1, 3), (1, 4), (2, 4), (3, 1), (4, 1), (4, 2)\}$
 $R_{v_1, v_3} := \{(1, 2), (1, 4), (2, 1), (2, 3), (3, 2), (3, 4), (4, 1), (4, 3)\}$
 ...

Constraint Networks
 Constraint networks
 Solution
 Normalized Constraint Networks
 Deduction
 Constraint Networks and Graphs
 Solving Constraint Networks

Example: Graph colorability

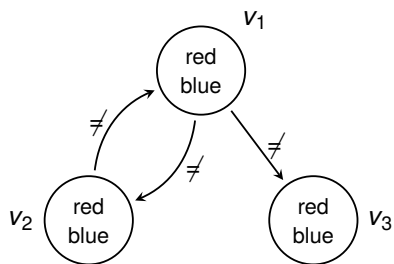


k -Colorability of a graph G can be represented as a constraint network of the following form:

$$V = \{v_i : v_i \text{ is a vertex in } G\}$$

$$D_i = \{1, \dots, k\} \quad (v_i \in V)$$

$$C = \{((v_i, v_j), \neq) : \{v_i, v_j\} \text{ is an edge of } G\}$$



Binary constraint networks can be represented by a **directed labeled graph** (or even: by an **undirected graph** if all constraints are

Constraint Networks
 Constraint networks
 Solution
 Normalized Constraint Networks
 Deduction
 Constraint Networks and Graphs
 Solving Constraint Networks

Solution of a constraint network



Definition

A **solution** of a constraint network $N = \langle V, D, C \rangle$ is a **(variable) assignment**

$$a: V \rightarrow \bigcup_{i: v_i \in V} D_i$$

such that:

- (a) $a(v_i) \in D_i$, for each $v_i \in V$,
- (b) $(a(x_1), \dots, a(x_m)) \in R$ for each constraint R_{x_1, \dots, x_m} in C .

N is called **satisfiable** if N has a solution.

$\text{sol}(N)$ denotes the set of all solutions of N . $\text{sol}(N)$ can also be written as:

$$\text{sol}(N) = \{(d_1, \dots, d_n) \in D_1 \times \dots \times D_n : \text{the assignment } v_1 \mapsto d_1, \dots, v_n \mapsto d_n \text{ defines a solution}\}$$

Constraint Networks
 Constraint networks
 Solution
 Normalized Constraint Networks
 Deduction
 Constraint Networks and Graphs
 Solving Constraint Networks

Instantiation, partial solution



Let $N = \langle V, D, C \rangle$ be a constraint network.

Definition

- An **instantiation** of a subset $V' \subseteq V$ is an assignment $a : V' \rightarrow \bigcup_{i: v_i \in V'} D_i$ with $a(v_i) \in D_i$.
- An instantiation a of V' is called **partial solution** if a satisfies each constraint R_S in C with $S \subseteq V'$. In this case a is called **locally consistent**.
- Shortcut notation: for an instantiation a of $V' = \{x_1, \dots, x_m\}$ and constraint R_S with scope $S \subseteq V'$, set

$$a[S] := (a(x_1), \dots, a(x_m)).$$

Hence, a solution is an instantiation of all variables in V that is locally consistent.

Constraint Networks
Constraint networks
Solution
Normalized Constraint Networks
Deduction
Constraint Networks and Graphs
Solving Constraint Networks

Instantiation, solution



Note:

- An instantiation of $V' \subseteq V$, a , is a partial solution (locally consistent) iff

$$a[S] \in R, \quad \text{for each constraint } R \text{ with scope } S \subseteq V'.$$
- Not every partial solution is part of a (full) solution, i.e., there may be partial solutions of a constraint network that cannot be extended to a solution. For the 4-queens problem, for example:

4	q			
3			q	
2				
1		q		
	v_1	v_2	v_3	v_4

Constraint Networks
Constraint networks
Solution
Normalized Constraint Networks
Deduction
Constraint Networks and Graphs
Solving Constraint Networks

Nogoods



Let $N = \langle V, D, C \rangle$ be a constraint network.

Definition

An instantiation a' of subset $V' \subseteq V$ is called a **nogood** (of N) if a' cannot be extended to a (full) solution of N , i.e., there exists no solution $a : V \rightarrow \bigcup_i D_i$ such that $a|_{V'} = a'$.

Instantiations that are no nogoods are sometimes called **consistent** or **globally consistent** (to emphasize the difference to locally consistent assignments).

Later, we will also introduce the notion of **globally consistent networks**.

Constraint Networks
Constraint networks
Solution
Normalized Constraint Networks
Deduction
Constraint Networks and Graphs
Solving Constraint Networks

Normalized constraint network



Let $N = \langle V, D, C \rangle$ be a constraint network.

Due to our definition it is possible that C contains constraints

$$R_{v_{i_1}, \dots, v_{j_k}} \quad \text{and} \quad S_{v_{j_1}, \dots, v_{j_k}}$$

where (j_1, \dots, j_k) is just a permutation of (i_1, \dots, i_k) .

Without changing the set of solutions, we can simplify the network by deleting $S_{v_{j_1}, \dots, v_{j_k}}$ from C and rewriting $R_{v_{i_1}, \dots, v_{j_k}}$ as follows:

$$R_{v_{i_1}, \dots, v_{j_k}} \leftarrow R_{v_{i_1}, \dots, v_{j_k}} \cap \pi_{v_{i_1}, \dots, v_{j_k}}(S_{v_{j_1}, \dots, v_{j_k}}).$$

Given a fixed order on the set of variables V , we can systematically delete-and-refine constraints. This results in a constraint network that contains **at most one constraint for each subset of variables**. Such a network is called a **normalized constraint network**.

Constraint Networks
Constraint networks
Solution
Normalized Constraint Networks
Deduction
Constraint Networks and Graphs
Solving Constraint Networks

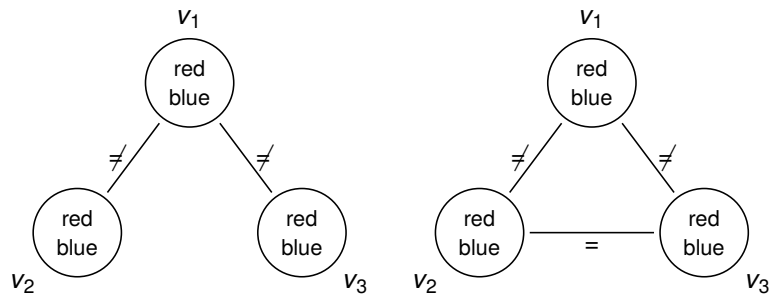
Equivalence

Let N and N' be constraint networks on the same set of variables and on the same domains for each variable.

Definition

N and N' are called **equivalent** if they have the same set of solutions.

Example:



- Constraint Networks
- Constraint networks
- Solution
- Normalized Constraint Networks
- Deduction
- Constraint Networks and Graphs
- Solving Constraint Networks

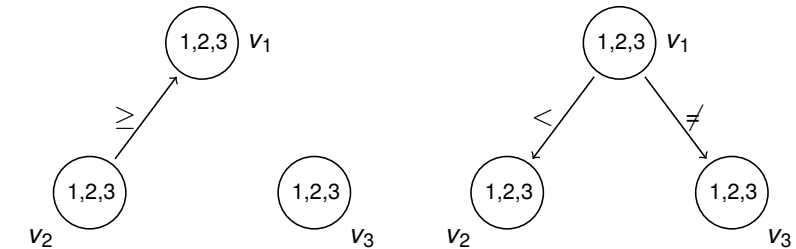
Tightness

Let N and N' be (normalized) constraint networks on the same set of variables and on the same domains for each variable.

Definition

N is **as tight as** N' if for each constraint R_S of N ,

- (a) N' has no constraint with the same scope as R_S , or
- (b) $R \subseteq \pi_S(R'_S)$, where R'_S is the constraint of N' with the same scope as R_S .



- Constraint Networks
- Constraint networks
- Solution
- Normalized Constraint Networks
- Deduction
- Constraint Networks and Graphs
- Solving Constraint Networks

- Clearly, if N' is as tight as N , then $\text{sol}(N') \subseteq \text{sol}(N)$.

Intersection of networks

Definition

The **intersection** of N and N' , $N \cap N'$, is the network defined by intersecting for each scope the constraints $R_S \in C$ and $R'_S \in C'$ with the same scope, i.e., modulo a suitable permutation of the constraint schemes,

$$R''_S := R_S \cap R'_S.$$

If for a scope S only one of the networks contains a constraint, then we set:

$$R''_S := R_S \quad (\text{or } := R'_S, \text{ resp.})$$

Lemma

If N and N' are equivalent networks, then $N \cap N'$ is equivalent to both networks and as tight as both networks.

- Constraint Networks
- Constraint networks
- Solution
- Normalized Constraint Networks
- Deduction
- Constraint Networks and Graphs
- Solving Constraint Networks

2 Constraint Networks and Graphs

- Primal Constraint Graphs
- Dual Constraint Graph
- Constraint Hypergraph

- Constraint Networks
- Constraint Networks and Graphs
- Primal Constraint Graphs
- Dual Constraint Graph
- Constraint Hypergraph
- Solving Constraint Networks

Primal constraint graphs



Let $N = \langle V, D, C \rangle$ be a (normalized) constraint network.

Definition

The **primal constraint graph** of a network $N = \langle V, D, C \rangle$ is the undirected graph

$$G_N := \langle V, E_N \rangle$$

where

$$\{u, v\} \in E_N \iff \{u, v\} \text{ is a subset of the scope of some constraint in } N.$$

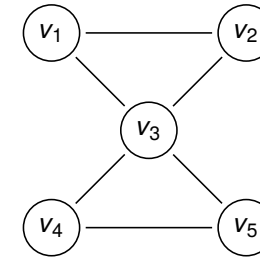
- Constraint Networks
- Constraint Networks and Graphs
- Primal Constraint Graphs
- Dual Constraint Graph
- Constraint Hypergraph
- Solving Constraint Networks

Primal constraint graph: Example



Consider a constraint network with variables v_1, \dots, v_5 and two ternary constraints R_{v_1, v_2, v_3} and S_{v_3, v_4, v_5} .

Then the primal constraint graph of the network has the form:



Absence of an edge between two variables/nodes means that there is no **explicit** constraint in which both variables participate.

- Constraint Networks
- Constraint Networks and Graphs
- Primal Constraint Graphs
- Dual Constraint Graph
- Constraint Hypergraph
- Solving Constraint Networks

Dual constraint graphs



Definition

The **dual constraint graph** of a constraint network $N = \langle V, D, C \rangle$ is the labeled graph

$$D_N := \langle V', E_N, I \rangle$$

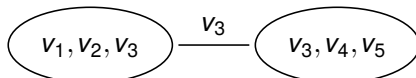
with

$$X \in V' \iff X \text{ is the scope of some constraint in } N$$

$$\{X, Y\} \in E_N \iff X \cap Y \neq \emptyset$$

$$I : E_N \rightarrow 2^V, \{X, Y\} \mapsto X \cap Y$$

In the example above, the dual constraint graph is:



- Constraint Networks
- Constraint Networks and Graphs
- Primal Constraint Graphs
- Dual Constraint Graph
- Constraint Hypergraph
- Solving Constraint Networks

Constraint hypergraph



Definition

The **constraint hypergraph** of a constraint network $N = \langle V, D, C \rangle$ is the hypergraph

$$H_N := \langle V, E_N \rangle$$

with

$$X \in E_N \iff X \text{ is the scope of some constraint in } N.$$

In the example above (constraint network with variables v_1, \dots, v_5 and two ternary constraints R_{v_1, v_2, v_3} and S_{v_3, v_4, v_5}) the hyperedges of the constraint hypergraph are:

$$E_N = \{ \{v_1, v_2, v_3\}, \{v_3, v_4, v_5\} \}.$$

- Constraint Networks
- Constraint Networks and Graphs
- Primal Constraint Graphs
- Dual Constraint Graph
- Constraint Hypergraph
- Solving Constraint Networks

3 Solving Constraint Networks



Constraint Networks
Constraint Networks and Graphs
Solving Constraint Networks

Simple solution strategy: Backtracking search



Constraint Networks
Constraint Networks and Graphs
Solving Constraint Networks

Backtracking: search systematically for locally consistent partial instantiations in a depth-first manner:

- **forward phase:** extend the current partial solution by assigning a consistent value to some new variable (if possible)
- **backward phase:** if no consistent instantiation for the current variable exists, return to the previous variable.

Backtracking algorithm



Constraint Networks
Constraint Networks and Graphs
Solving Constraint Networks

Backtracking(N, a):

Input: a constraint network $N = \langle V, D, C \rangle$ and a partial assignment a of N
(e.g., the empty instantiation $a = \{ \}$)

Output: a solution of N or “inconsistent”

if a is not locally consistent with N :

return “inconsistent”

if a is defined for all variables in V :

return a

select **some variable** v_i for which a is not defined

for each value x from D_i :

$a' := a \cup \{v_i \mapsto x\}$

$a'' \leftarrow \text{Backtracking}(N, a')$

if a'' is not “inconsistent”:

return a''

return “inconsistent”

Literature



Constraint Networks
Constraint Networks and Graphs
Solving Constraint Networks



Rina Dechter.

Constraint Processing,

Chapter 2, Morgan Kaufmann, 2003