

Constraint Programming

Matteo Zavatteri

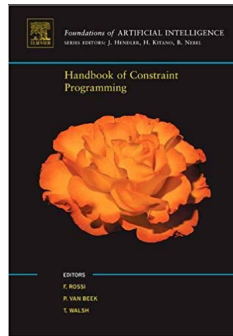
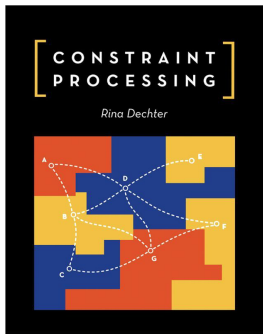
Math Decisions 2020

Outline



- ① Constraint networks
- ② Global constraints
- ③ More expressiveness
- ④ Modeling CSPs

Reference books



More online:

<http://www.constraint-programming.com/people/regin/papers/global.pdf>

<https://www.minizinc.org> (have a look at *MiniZinc Handbook*)

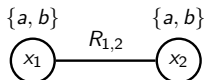
Constraint Networks

Formal definition

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

- ❶ $X = \{x_1, \dots, x_n\}$ is a finite set of (*decision*) *variables*
- ❷ $D = \{D_1, \dots, D_n\}$ is a set of associated domains.
- ❸ C is a finite set of *constraints*. Each constraint is a relation $R_{i, \dots, k}$ (defined over the set of variables $\{x_i, \dots, x_k\}$ such that $R_{i, \dots, k} \subseteq D_i \times \dots \times D_k$).

To ease notation, scopes and tuples are “ordered” with respect to variable indexes.



Formal specification

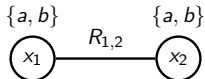
- $X = \{x_1, x_2\}$, $D = \{D_1, D_2\}$
- $D_1 = D_2 = \{a, b\}$
- $C = \{R_{1,2}\}$, where $R_{1,2} = \{(a, b), (b, a)\}$ (i.e., $x_1 \neq x_2$)

Consistency

Consistency

A constraint network is consistent if there exists a solution. That is, if every variable can be assigned a value from its domain such that all constraints are eventually satisfied.

Given a solution $x_1 = v_1, \dots, x_n = v_n$, a constraint $R_{p,\dots,z}$ is satisfied, if $(v_p, \dots, v_z) \in R$, where $x_p = v_p, \dots, x_z = v_z$ are in the solution.



Solution: $x_1 = a, x_2 = b$

A few problems associated to constraint networks

Given a constraint network N , we might address the following problems:

Decision problems:

- Is N consistent/inconsistent?
- Does N admit at least/at most/exactly k different solutions?
- Is there an assignment satisfying at least k constraints?

Search problems:

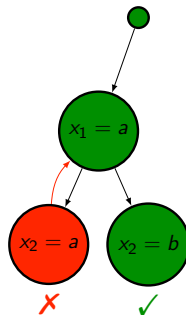
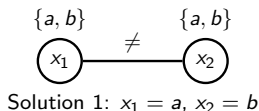
- Find a consistent assignment
- Find 2,3,etc different consistent assignments
- Find all consistent assignments
- How many different consistent assignments does N admit?
- Find an assignment maximizing the number of satisfied constraints

In this part, we will only address search problems for which we need to find a fixed number of solutions.

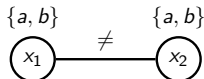
Search for 1 solution: backtracking

Assume a recursive algorithm that assigns variables according to the order of their indexes.

The algorithm stops *as soon as* it finds a solution

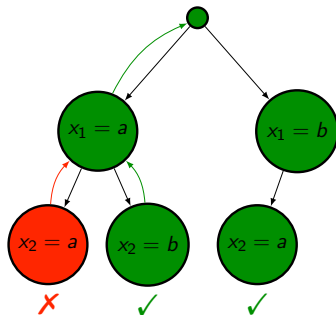


Search for 2 solutions: keep searching up to the 2nd

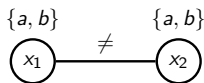


Solution 1: $x_1 = a, x_2 = b$

Solution 2: $x_1 = b, x_2 = a$

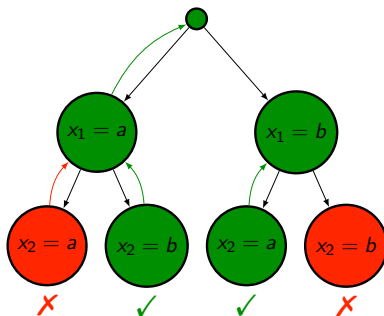


Search for all solutions: keep searching up to the end



Solution 1: $x_1 = a, x_2 = b$

Solution 2: $x_1 = b, x_2 = a$



Filtering domains: node consistency

Node consistency

A variable x_i is node consistent if for each $v \in D_i$ we have that $(v) \in R_i$.

$\{a, b, c\}$



R_i

$$R_i = \{(a), (c)\}$$

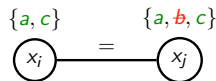
- x_i is not node consistent as $b \notin R_i$
- Removing b from D_i makes x_i node consistent

Rationale: every solution must satisfy R_i and $x_i = b$ just doesn't.

Filtering domains: arc consistency

Arc consistency

A pair of different variables x_i, x_j is arc consistent if for each $v_i \in D_i$ there exists $v_j \in D_j$ such that $(v_i, v_j) \in R_{ij}$.



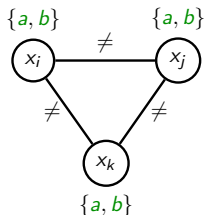
- x_i, x_j are not arc consistent. $(a, b) \notin R_{i,j}$, $(c, b) \notin R_{i,j}$
- Removing b from D_j makes x_i, x_j arc consistent.

Rationale: every solution must satisfy $R_{i,j}$ and $x_j = b$ (whatever x_i) just doesn't.

Filtering domains: path consistency

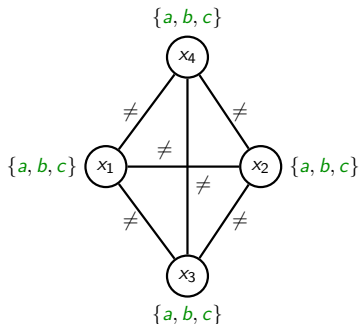
Path consistency

A pair of variables x_i, x_j is path consistent with another variable x_k ($x_i \neq x_j \neq x_k$) if for each $v_i \in D_i, v_j \in D_j$ with $(v_i, v_j) \in R_{ij}$, there exists $v_k \in D_k$ such that $(v_i, v_k) \in R_{ik}$ and $(v_j, v_k) \in R_{jk}$



- Arc consistent!
- Not path consistent. $x_i = a, x_j = b$ cannot be extended to any $x_k = v_k$ where $v_k \in \{a, b\}$.
- The network is actually inconsistent.

Path consistency is not enough!



- Path consistent!
- Yet, the network is actually inconsistent.
- All variables must get different values. Four variables. Three values.
- Examples like this extend to networks with n variables, $n - 1$ values in each domain and a “ \neq ” constraint between any pair of distinct variables.
- Enforcing consistency on n variables says nothing for $n + 1$ variables.

Node, arc and path consistency are *pruning techniques* to rule out (even many) values from domains. But eventually, we still need to search.

Global constraints

Global constraints = compact constraints

- they encapsulate several constraints in a single one
- they avoid writing an explicit relation of many, many tuples
- they typically involve several variables
- they allow for the specification of “high level” constraints

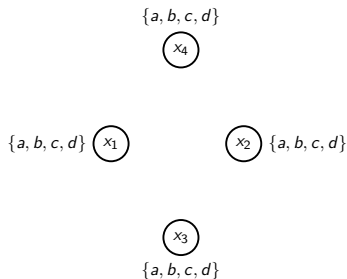
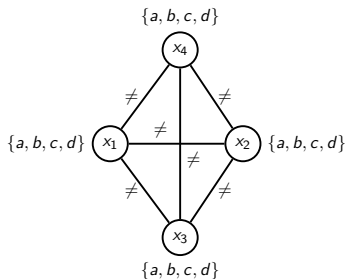
Examples:

- `all_different(x_1, \dots, x_n)`
- ...

all_different

All different

A solution $x_1 = v_1, \dots, x_n = v_n$ to a constraint network satisfies an $\text{all_different}(x_i, \dots, x_j)$ iff $v_i \neq \dots \neq v_j$.



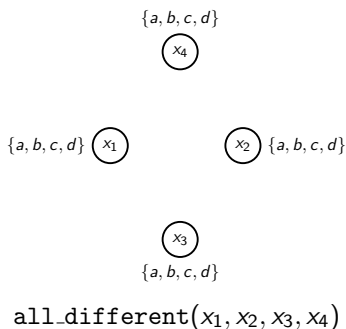
$\text{all_different}(x_1, x_2, x_3, x_4)$

$x_1 = a, x_2 = b, x_3 = c, x_4 = d$

all_different, more formally

A possible formal definition

`all_different`(x_1, \dots, x_n) is equivalent to a relation $R_{1,\dots,n}$ such that for each tuple $(v_1, \dots, v_n) \in R$, it holds that $|\{v_i \mid v_i \in (v_1, \dots, v_n)\}| = n$.



$R_{1,2,3,4} = \{(a, b, c, d), (a, b, d, c),$
 $(a, c, b, d), (a, c, d, b), (a, d, b, c),$
 $(a, d, c, b), (b, a, c, d), (b, a, d, c),$
 $(b, c, a, d), (b, c, d, a), (b, d, a, c),$
 $(b, d, c, a), (c, a, b, d), (c, a, d, b),$
 $(c, b, a, d), (c, b, d, a), (c, d, a, b),$
 $(c, d, b, a), (d, a, b, c), (d, a, c, b),$
 $(d, b, a, c), (d, b, c, a), (d, c, a, b),$
 $(d, c, b, a)\}$

$$x_1 = c, x_2 = a, x_3 = b, x_4 = d$$

Boosting expressiveness maintaining complexity

Main complexity result

Deciding consistency of (classic) constraint networks is NP-complete.

- it is easy to see that the problem remains NP-complete even if we add global constraints or we turn a set of constraints into a boolean formula where global constraints and relations play the role of boolean atoms (provided that, given a solution, the satisfaction of each atom can be checked in polynomial time).

$F ::= R_{i,\dots,k} \mid \text{global_constraint}(\dots) \mid \neg F \mid (F) \mid F \square F$ where
 $\square \in \{\wedge, \vee, \Rightarrow, \Leftrightarrow, \dots\}$

$\{a, b, c, d\}$



$\{a, b, c, d\}$



Let $x_i = v_j$ be a short for $R_i = \{(v_j)\}$ (i.e., a further constraint language improvement!). The formula:

$\text{all_different}(x_1, x_2, x_3) \wedge (x_1 = a \vee x_3 = c) \wedge x_2 = d \wedge (x_1 = a \Rightarrow x_2 = c)$



$\{a, b, c, d\}$

is satisfied by the solution $x_1 = b, x_2 = d, x_3 = c$

The solution (certificate of yes) can still be checked in polynomial time!

Humanizing relations by means of formulae

Consider the following constraint language:

$$F ::= x = v \mid (F) \mid F \wedge F \mid F \vee F$$

Wouldn't it be enough to encode a set of constraints $R_{i,\dots,z}$? (yes!)

Consider a constraint network $N = \langle X, D, C \rangle$ where:

- $X = \{x_1, x_2, x_3\}$
- $D_1 = D_2 = D_3 = \{a, b\}$
- $C = \{R_2, R_{13}, R_{123}\}$, where $R_2 = \{(b)\}$, $R_{13} = \{(a, a), (b, a), (b, b)\}$, $R_{123} = \{(a, b, a), (b, a, b)\}$

C can be encoded in a (DNF) formula $F \equiv \underbrace{F_2}_{R_2} \wedge \underbrace{F_{13}}_{R_{13}} \wedge \underbrace{F_{123}}_{R_{123}}$, where:

- $F_2 \equiv (x_2 = b)$
- $F_{13} \equiv ((x_1 = a \wedge x_3 = a) \vee (x_1 = b \wedge x_3 = a) \vee (x_1 = b \wedge x_3 = b))$
- $F_{123} \equiv ((x_1 = a \wedge x_2 = b \wedge x_3 = a) \vee (x_1 = b \wedge x_2 = a \wedge x_3 = b))$

In general $R_{i,\dots,z}$ can be encoded in $F \equiv (\bigvee_{(v_i,\dots,v_z) \in R_{i,\dots,z}} (x_i = v_i \wedge \dots \wedge x_z = v_z))$

Modeling constraint satisfaction problems (CSP)

In what follows, we will:

- ➊ start with the definition of some problem in natural and formal language
- ➋ model it in the input language of MiniZinc
- ➌ push a button to search for one (or more) solution(s)

In this order. This is what we are going to do.

MiniZinc

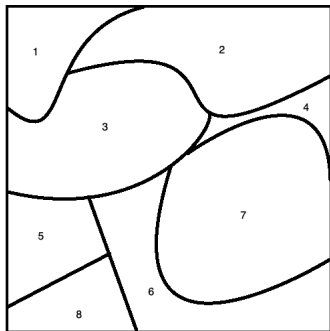


<https://www.minizinc.org>

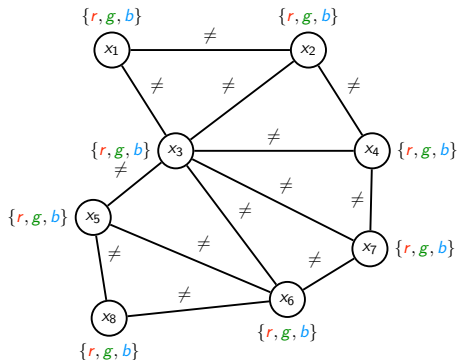
- MiniZinc is a free and open-source constraint modeling language that allows you to write models that are compiled into FlatZinc: an input language that is understood by a wide range of solvers.
- MiniZinc is developed at item Monash University MonashUniversity in collaboration with Data61 Decision Sciences <https://research.csiro.au/data61/tag/decision-sciences/> and the University of Melbourne <https://unimelb.edu.au>.
- MiniZinc is available for Windows, Linux and MacOS. Have a look at <https://www.minizinc.org/software.html>, download and install it on your computer.

Modeling CSPs: Map coloring

Can you color this map by using red, green and blue so that any two adjacent regions have different colors?

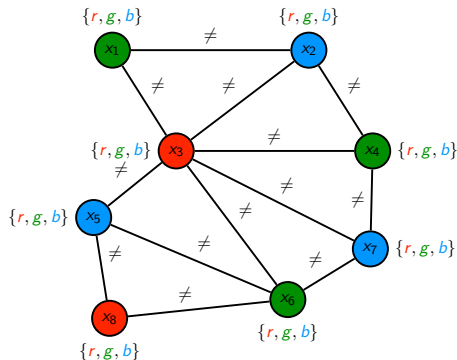


Constraint Network formulation

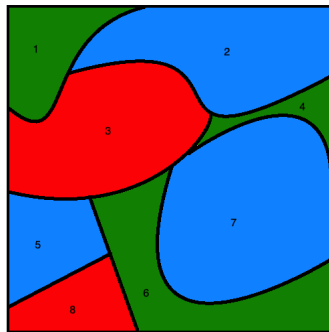


Modeling CSPs: Map coloring

Solution to the corresponding constraint network



Back to the map!

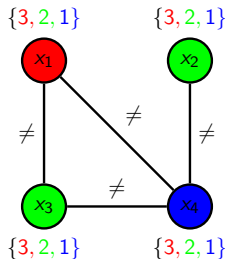


Graph K -Coloring Problem

Input. A graph $G = (V, E)$ and a positive integer K .

Output. yes iff there exists $f: V \rightarrow \{1, \dots, K\}$ s.t. $f(u) \neq f(v)$ for each $(u, v) \in E$

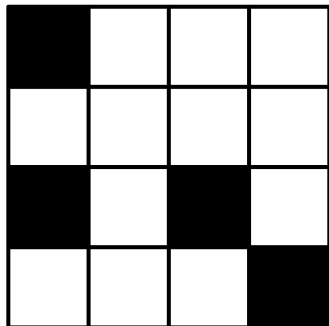
Example. $G = (V, E)$, where $V = \{x_1, x_2, x_3, x_4\}$ and $E = \{(x_1, x_3), (x_1, x_4), (x_3, x_4), (x_2, x_4)\}$ and $K = 3$.



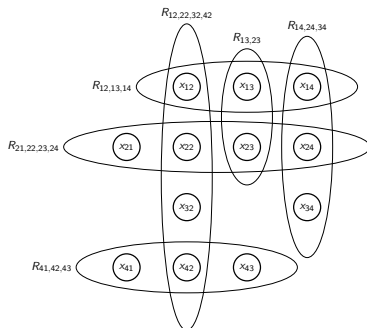
$$f(x_1) = r, f(x_2) = g, f(x_3) = g, f(x_4) = b.$$

Modeling CSPs: Crossword puzzle

Can you fill the crossword puzzle, by only using the words CAT, LATE, TED, TEA, CAKE, AT (repetitions allowed)?



Constraint Network formulation



For each x_{ij} , $D_{ij} = \{A, C, D, E, K, L, T\}$

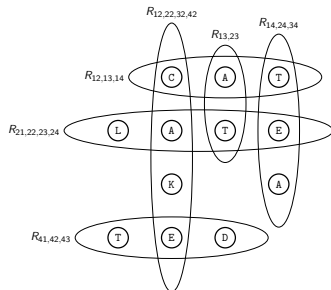
$$R_{12,13,14} = \{(C, A, T), (T, E, D), (T, E, A)\}, R_{21,22,23,24} = \{(L, A, T, E), (C, A, K, E)\},$$

$$R_{41,42,43} = \{(C, A, T), (T, E, D), (T, E, A)\}, R_{12,22,32,42} = \{(L, A, T, E), (C, A, K, E)\},$$

$$R_{13,23} = \{(A, T)\}, R_{14,24,34} = \{(C, A, T), (T, E, D), (T, E, A)\}$$

Modeling CSPs: Crossword puzzle

Solution to the corresponding constraint network



Back to crossword puzzle

	C	A	T
L	A	T	E
	K		A
T	E	D	

For each x_{ij} , $D_{ij} = \{A, C, D, E, K, L, T\}$

$$\begin{aligned}
 R_{12,13,14} &= \{(C, A, T), (T, E, D), (T, E, A)\}, & R_{21,22,23,24} &= \{(L, A, T, E), (C, A, K, E)\}, \\
 R_{41,42,43} &= \{(C, A, T), (T, E, D), (T, E, A)\}, & R_{12,22,32,42} &= \{(L, A, T, E), (C, A, K, E)\}, \\
 R_{13,23} &= \{(A, T)\}, & R_{14,24,34} &= \{(C, A, T), (T, E, D), (T, E, A)\}
 \end{aligned}$$

Sudoku

Input. A 9x9 grid in which each cell (i,j) must be filled with digits from 1 to 9. Some cells are prefilled (coherently with what the solution should look like).

Output. A filling of all empty cells of the grid such that each digit appears exactly once in each row, each column and each 3x3 subsquare.

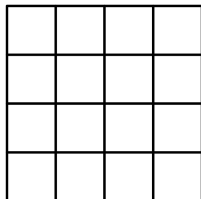
Example.

	5	9	8				7	
				1				
3					2	5		8
			6	2				1
	8	2		3		4		
		6					3	
9				7				3
				9			2	
	4	8			6	7		

6	5	9	8	4	3	1	7	2
8	2	7	5	1	9	3	4	6
3	1	4	7	6	2	5	9	8
4	7	3	6	2	5	9	8	1
5	8	2	9	3	1	4	6	7
1	9	6	4	8	7	2	3	5
9	6	1	2	7	4	8	5	3
7	3	5	1	9	8	6	2	4
2	4	8	3	5	6	7	1	9

Modeling CSPs: 4-queens problem

Can you place 4 queens on a 4×4 board such that no queen attacks any other? I.e., no row, column, nor diagonal sharing.

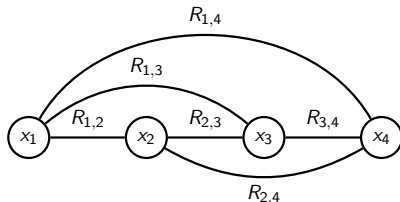


Constraint Network formulation

j variables, one for each column.

$D_i = \{1, 2, 3, 4\}$ for $i = 1, \dots, 4$ (rows).

$x_j = i$ means that there is a queen in (i, j)



$$R_{1,2} = \{(1, 3), (1, 4), (2, 4), (3, 1), (4, 1), (4, 2)\},$$

$$R_{1,3} = \{(1, 2), (1, 4), (2, 1), (2, 3), (3, 2), (3, 4), (4, 1), (4, 2), (4, 3)\},$$

$$R_{1,4} = \{(1, 2), (1, 3), (2, 1), (2, 3), (2, 4), (3, 1), (3, 2), (3, 4), (4, 2), (4, 3)\},$$

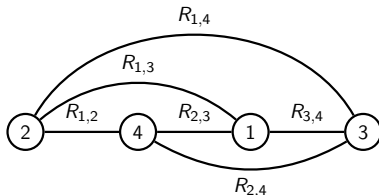
$$R_{2,3} = \{(1, 3), (1, 4), (2, 4), (3, 1), (4, 1), (4, 2)\}$$

$$R_{2,4} = \{(1, 2), (1, 4), (2, 1), (2, 3), (3, 2), (3, 4), (4, 1), (4, 3)\}$$

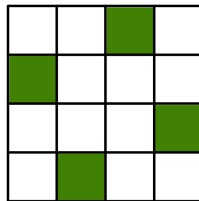
$$R_{3,4} = \{(1, 3), (1, 4), (2, 4), (3, 1), (4, 1), (4, 2)\}$$

Modeling CSPs: 4-queens problem

Solution to the constraint network



Back to the board.



$$R_{1,2} = \{(1, 3), (1, 4), (2, 4), (3, 1), (4, 1), (4, 2)\},$$

$$R_{1,3} = \{(1, 2), (1, 4), (2, 1), (2, 3), (3, 2), (3, 4), (4, 1), (4, 2), (4, 3)\},$$

$$R_{1,4} = \{(1, 2), (1, 3), (2, 1), (2, 3), (2, 4), (3, 1), (3, 2), (3, 4), (4, 2), (4, 3)\},$$

$$R_{2,3} = \{(1, 3), (1, 4), (2, 4), (3, 1), (4, 1), (4, 2)\}$$

$$R_{2,4} = \{(1, 2), (1, 4), (2, 1), (2, 3), (3, 2), (3, 4), (4, 1), (4, 3)\}$$

$$R_{3,4} = \{(1, 3), (1, 4), (2, 4), (3, 1), (4, 1), (4, 2)\}$$

Interpretation of constraints: each pair $(i, j) \in R_{a,b}$ means that the placement of two queens in positions (i, a) and (j, b) is permitted.

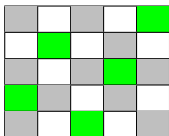
N queens

Input. An $N \times N$ board.

Output. A placement of N queens such that no queen can attack any other in one move.

Note. A queen can attack another queen if the two queens are on the same row, column or diagonal.

Example. $N = 5$



(4, 1), (2, 2), (5, 3), (3, 4),
(1, 5)

Two queens are on the same diagonal if one of the following conditions hold:

- The row number plus the column number for each of the two queens are equal
- The row number minus the column number for each of the two queens are equal

Modeling CSPs: (very) simple activity scheduling

4 tasks A, B, C, D to be scheduled.
Suppose any task is scheduled at a specific hour and takes that hour to be executed

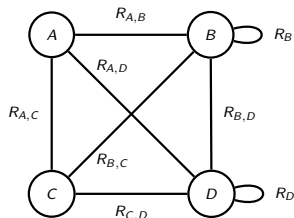
Constraints:

- B is either the 1st or the 4th to be executed
- D is either the 3rd or the 4th to be executed
- One task at a time (no 2 tasks are executed simultaneously)
- B before A , A before D , C after B

Constraint network formulation

Assume A, B, C, D resemble the indexes 1, 2, 3, 4.

$D_I = \{1, 2, 3, 4\}$ for $I = A, B, C, D$.



$$R_B = \{(1), (4)\}, R_D = \{(3), (4)\}$$

$$R_{I,J} = \{(1, 2), (1, 3), (1, 4), (2, 1), (2, 3), (2, 4), (3, 1), (3, 2), (3, 4)\} \text{ for } I, J = A, B, C, D \text{ with } I \neq J$$

$$R_{A,B} = \{(2, 1), (3, 1), (3, 2), (4, 3), (4, 2), (4, 1)\}$$

$$R_{A,D} = \{(1, 2), (1, 3), (1, 4), (2, 3), (2, 4), (3, 4)\}$$

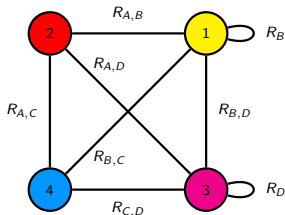
$$R_{B,C} = \{(1, 2), (1, 3), (1, 4), (2, 3), (2, 4), (3, 4)\}$$

Modeling CSPs: (very) simple activity scheduling

Solution to the constraint

A, B, C, D resemble the indexes 1, 2, 3, 4.

$D_I = \{1, 2, 3, 4\}$ for $I = A, B, C, D$.



$R_B = \{(1), (4)\}$, $R_D = \{(3), (4)\}$

$R_{I,J} = \{(1, 2), (1, 3), (1, 4), (2, 1), (2, 3), (2, 4), (3, 1), (3, 2), (3, 4)\}$ for $I, J = A, B, C, D$ with $I \neq J$ (e.g., $R_{A,B}$)

$R_{A,B} = \{(2, 1), (3, 1), (3, 2), (4, 3), (4, 2), (4, 1)\}$

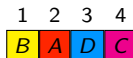
$R_{A,D} = \{(1, 2), (1, 3), (1, 4), (2, 3), (2, 4), (3, 4)\}$

$R_{B,C} = \{(1, 2), (1, 3), (1, 4), (2, 3), (2, 4), (3, 4)\}$

Back to the scheduling problem

Constraints:

- B is either the 1st or the 4th to be executed
- D is either the 3rd or the 4th to be executed
- One task at a time (no 2 tasks are executed simultaneously)
- B before A , A before D , C after B



Modeling CSPs: 3SAT

SAT: A boolean formula φ in CNF (=conjunctive normal form) over a set of variables x_1, \dots, x_n

A CNF is a conjunction of clauses, where each clause is a disjunction of literals. E.g., $(x_1 \vee x_2) \wedge (\neg x_3 \vee x_4 \vee \neg x_5) \dots$

3SAT: A SAT instance in which each clause has exactly 3 literals.

Consider

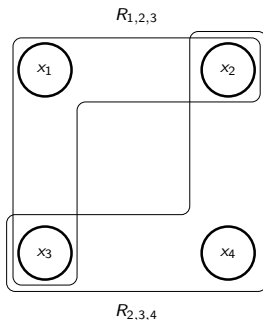
$$\varphi \equiv \underbrace{(x_1 \vee \neg x_2 \vee x_3)}_{\text{First clause}} \wedge \underbrace{(x_2 \vee \neg x_3 \vee \neg x_4)}_{\text{Second clause}}$$

$$R_{1,2,3} = \{(0, 0, 0), (0, 0, 1), (0, 1, 1), (1, 0, 0), (1, 0, 1), (1, 1, 0), (1, 1, 1)\}$$

$$R_{2,3,4} = \{(0, 0, 0), (0, 0, 1), (0, 1, 0), (1, 0, 0), (1, 0, 1), (1, 1, 0), (1, 1, 1)\}$$

Constraint Network formulation

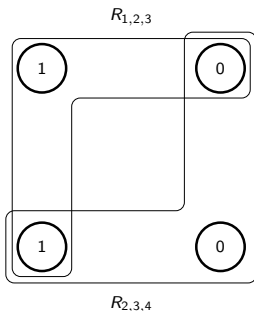
$$D_i = \{0, 1\} \text{ for } i = 1, \dots, 4.$$



Modeling CSPs: 3SAT

Constraint Network formulation

$D_i = \{0, 1\}$ for $i = 1, \dots, 4$.



$R_{1,2,3} = \{(0, 0, 0), (0, 0, 1), (0, 1, 1), (1, 0, 0), (1, 0, 1), (1, 1, 0), (1, 1, 1)\}$

$R_{2,3,4} = \{(0, 0, 0), (0, 0, 1), (0, 1, 0), (1, 0, 0), (1, 0, 1), (1, 1, 0), (1, 1, 1)\}$

Back to 3SAT

$$\varphi \equiv \underbrace{(x_1 \vee \neg x_2 \vee x_3)}_{\text{First clause}} \wedge \underbrace{(x_2 \vee \neg x_3 \vee \neg x_4)}_{\text{Second clause}}$$

Model: $x_1 = 1, x_2 = 0, x_3 = 1, x_4 = 0$,

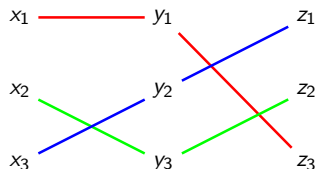
Three Dimensional Matching

Input. 3 disjoint sets X, Y, Z each of size n and a set $T \subseteq X \times Y \times Z$.

Output. yes iff there exists $T_1 \subseteq T$ such that $|T_1| = n$ and for each $t \in X \cup Y \cup Z$, t is contained exactly in one of the triples of T_1 .

Example $X = \{x_1, x_2, x_3\}$, $Y = \{y_1, y_2, y_3\}$, $Z = \{z_1, z_2, z_3\}$

$T = \{(x_1, y_1, z_3), (x_2, y_1, z_1), (x_2, y_3, z_2), (x_3, y_1, z_3), (x_3, y_2, z_1)\}$



$T_1 = \{(x_1, y_1, z_3), (x_2, y_3, z_2), (x_3, y_2, z_1)\}$

Knapsack Problem

Input. A finite set I of items where for each $i \in I$, $w(i) \in \mathbb{Z}^+$ is the weight and $v(i) \in \mathbb{Z}^+$ is the value of item i , and two positive integers W and V .

Output. yes iff there exists $I_1 \subseteq I$ such that $\sum_{i \in I_1} w(i) \leq W$ and $\sum_{i \in I_1} v(i) \geq V$.

Example: $W = 17$, $V = 32$

Item	Weight	Value
1	1	7
2	8	7
3	3	8
4	6	10
5	5	7
6	8	1
7	7	4
8	2	3
9	9	5
10	10	1

Set Packing

Input. A collection C of finite sets and a positive integer $K \leq |C|$ **Output.** yes iff C contains at least K mutually disjoint sets.

Example $C = \{\{2, 4\}, \{1, 3, 5\}, \{1, 2, 4\}, \{1, 2, 5\}, \{2, 3\}, \{6, 8\}, \{1, 3, 9\}, \{10, 7\}, \{10, 2, 3\}, \{3, 5, 8\}\}$, $K = 4$;

Set Splitting

Input. A collection C of subsets of a finite set S .

Output. yes iff there exists a partition of S in two subsets S_1 and S_2 such that for each $C_i \in C$, C_i is neither a subset of S_1 nor S_2 .

Example $S = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}$,

$C = \{\{1, 2, 4\}, \{1, 2, 5\}, \{2, 3\}, \{10, 7\}, \{10, 2, 3\}, \{3, 5, 8\}\}$

$S_1 = \{3, 4, 5, 10\}$, $S_2 = \{1, 2, 6, 7, 8, 9\}$

Subset Sum

Input. A finite set X of elements where for each $x \in X$, $s(x) \in \mathbb{Z}$, and an integer S .

Output. yes iff there exists $X_1 \subseteq X$ such that $\sum_{x \in X_1} s(x) = S$.

Example $S = 13$

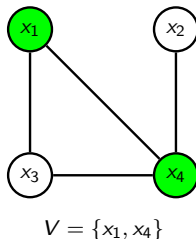
S	s(x)
1	-1
2	-8
3	3
4	-6
5	-5
6	8
7	-7
8	2
9	-9
10	10

Vertex Cover

Input. A graph $G = (V, E)$ and a positive integer K .

Output. yes iff there exists $V_1 \subseteq V$ such that $|V_1| \leq K$ and for each $(u, v) \in E$, u or v is in V_1 (might also be both)

Example. $G = (V, E)$, where $V = \{x_1, x_2, x_3, x_4\}$,
 $E = \{(x_1, x_3), (x_1, x_4), (x_3, x_4), (x_4, x_2)\}$ and $K = 2$.

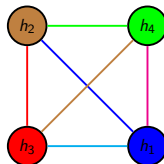
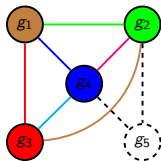


SubGraph Isomorphism

Input. Two (undirected) graphs $G = (V_G, E_G)$ and $H = (V_H, E_H)$

Output. yes iff there exists a graph $I = (V_I, E_I)$ and an injection $f : V_H \rightarrow V_I$ such that I is a subgraph of G (i.e., $V_I \subseteq V_G$ and $E_I \subseteq E_G$), $|V_I| = |V_H|$, $|E_I| = |E_H|$, $(f(v), f(u)) \in E_I$ iff $(u, v) \in E_H$.

Example.



$f(h_1) = g_4$, $f(h_2) = g_3$, $f(h_3) = g_2$, $f(h_4) = g_1$

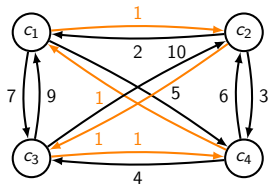
Traveling Salesman Problem

Input. A set C of N cities, a positive integer B and a distance matrix D $N \times N$ where each $D[i, j]$ is a positive integer containing the distance to go from city i to city j .

Output. yes iff there exists an injection $f: \{1, \dots, N\} \rightarrow C$ such that

$$(\sum_{i=1}^{N-1} D[f(i), f(i+1)]) + D[f(N), f(1)] \leq B$$

Example. $C = \{c_1, c_2, c_3, c_4\}$, $B = 5$ and D (see below).



$$D = \begin{bmatrix} 0 & 1 & 7 & 5 \\ 2 & 0 & 1 & 3 \\ 9 & 10 & 0 & 1 \\ 1 & 6 & 4 & 0 \end{bmatrix}$$

$$f(1) = c_2, f(2) = c_3, f(3) = c_4, f(4) = c_1$$

$$(c_2 \rightarrow c_3 \rightarrow c_4 \rightarrow c_1 \rightarrow c_2)$$