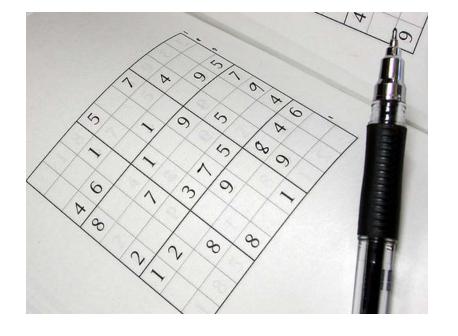
Constraint Programming

Mikael Z. Lagerkvist

Tomologic

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- Introduction
- Solving Sudoku with CP
- Constraint programming basics
- Examples
- 5 Constraint programming in perspective
- Summary

Who am I?

- Mikael Zayenz Lagerkvist
- Basic education at KTH 2000-2005
 - Datateknik
- PhD studies at KTH 2005-2010
 - Research in constraint programming systems
 - One of three core developers for Gecode, fast and well-known Constraint Programming (CP) system.
 - http://www.gecode.org
- Senior developer R&D at Tomologic
 - Optimization systems for sheet metal cutting
 - Constraint programming for some tasks







Tomologic

- Mostly custom algorithms and heuristics
- Part of system implemented using CP at one point
 - Laser Cutting Path Planning Using CP
 Principles and Practice of Constraint Programming 2013
 M. Z. Lagerkvist, M. Nordkvist, M. Rattfeldt
- Some sub-problems solved using CP
 - Ordering problems with side constraints
 - Some covering problems

Sudoku - The Rules

- Each square gets one value between 1 and 9
- Each row has all values different
- Each column has all values different.
- Each square has all values different

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

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

$$X = \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$$

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

$$X = \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$$

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

$$X = \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$$

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

$$X = \{2, 3, 6, 7, 8, 9\}$$

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

$$X = \{2, 3, 6, 7, 8, 9\}$$

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

$$X = \{2, 3, 6, 7, 8, 9\}$$

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

$$X = \{3, 6, 7\}$$

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

$$X = \{3, 6, 7\}$$

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

$$X = \{3, 6, 7\}$$

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

$$X = \{6\}$$

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

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

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

Sudoku - Solving with CP

- Solving Sudoku using CP
- Defining the variables
- Defining the constraints
- I will use the MiniZinc modelling language
 - ▶ http://www.minizinc.org/
 - High level modelling for CP
 - Model in MiniZinc solvable using many different systems

Sudoku - Defining the variables

```
array[1...9,1...9] of var 1...9:
   puzzle = [|
        _, _, _, _, 3, _, 6, _|
        _, _, _, _, _, _, 1, _|
        _, 9, 7, 5, _, _, _, 8, _|
        _, _, _, _, 9, _, 2, _, _|
        _, _, 8, _, 7, _, 4, _, _|
        _, _, 3, _, 6, _, _, _, _|
        _, 1, _, _, 2, 8, 9, _|
        _, 4, _, _, _, _, _, _, _|
        _, 5, _, 1, _, _, _, _, _|
    ];
```

Sudoku - Rules for rows and columns

```
% In all columns, all values different
constraint forall (col in 1..9) (
           all different (row in 1..9)
              (puzzle[row, col])
       );
% In all rows, all values different
constraint forall (row in 1..9) (
           all_different (col in 1..9)
              (puzzle[row, col])
       );
```

Sudoku - Rules for squares

Sudoku - Search and output a solution

```
solve satisfy;
output [ show(puzzle[i,j]) ++
         if j = 9 then "\n"
         else " "
         endif
         | i,j in 1..9 ];
```

Sudoku - Full program

```
include "globals.mzn";
array[1..9,1..9] of var 1..9 : puzzle = [|
        _, _, _, _, 3, _, 6, _|
        _, _, _, _, _, _, _, 1, _|
        _, 9, 7, 5, _, _, _, 8, _|
        _, _, _, _, 9, _, 2, _, _|
        _, _, 8, _, 7, _, 4, _, _|
        _, _, 3, _, 6, _, _, _, _|
        _, 1, _, _, 2, 8, 9, _|
        _, 4, _, _, _, _, _, _, _,
        _, 5, _, 1, _, _, _, _, _|
    11:
% In all columns, all values different
constraint forall (col in 1..9) (
           all different (row in 1..9) (puzzle[row, col]) :: domain
       );
% In all rows, all values different
constraint forall (row in 1..9) (
           all_different (col in 1..9) (puzzle[row, col]) :: domain
       ):
% In all squares, all values different
constraint forall (row.col in {1.4.7}) (
           all_different (i,j in 0..2) (puzzle[row+i, col+j]) :: domain
       );
solve satisfy:
output [ show(puzzle[i,j]) ++ if j = 9 then "\n" else " " endif
       | i, i in 1..9];
```

Sudoku - Solution

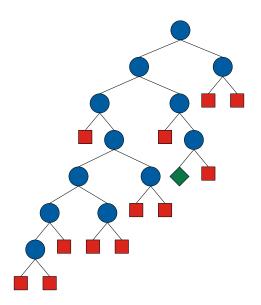
```
mzn-gecode -a -s sudoku.mzn
  1 8 5 9 2 3 7 6 4
  2 3 4 6 8 7 5 1 9
  6 9 7 5 1 4 3 8 2
  471398256
  9 6 8 2 7 5 4 3 1
  5 2 3 4 6 1 9 7 8
  3 1 6 7 4 2 8 9 5
  7 4 9 8 5 6 1 2 3
  8 5 2 1 3 9 6 4 7
```

. . .

Sudoku - Statistics

```
%%
                      0.001 (1.412 \text{ ms})
    runtime:
%%
                      0.000 (0.411 \text{ ms})
   solvetime:
%%
    solutions:
%%
    variables:
                      81
%%
    propagators:
                      27
%%
                      379
    propagations:
%%
    nodes:
                      12
%% failures:
                      5
%%
    peak depth:
                      4
%%
    peak memory:
                      100 KB
```

Sudoku - Search tree



What is Constraint programming?

- A way to model combinatorial (optimization) problems
- Systems for solving such problems
- A programming paradigm
- Theoretical model used in complexity

Constraint programming for modelling

- Modelling combinatorial (optimization) problems
- Problems are typically complex (NP-hard)
- Language for describing models and instances
- Solving using different techniques
 - dedicated constraint programming systems most common

Constraint programming systems

- Uses modeled structure for smart search
- Problems are typically complex (NP-hard)
 A CP system is no silver bullet
- Often implemented as libraries
 - Commercial: IBM CP Optimizer (C++), Xpress Kalis,
 Opturion CPX (C++), Sicstus Prolog (C), ...
 - ► Free: Gecode (C++), Choco (Java), OscaR (Scala), Google or-tools (C++), Minion (C++), Jacop (Java),

. . .

Constraint programming as a paradigm

- Constraint Logic Programming
 - Mix of libraries and language
 - Search through Prolog search
 - Sicstus Prolog, ECLiPSe, BProlog, ...
- Constraint Handling Rules
 - Language for expressing constraints
- Other languages such as Mozart/Oz
 - Constraints embedded into base language
 - For example, used as synchronization mechanism between threads

Constraint programming as theoretical model

- Using theoretical models of constraint problems for complexity research
- Not interesting for solving practical problems
- Not my area

Constraint programming in my view

- Both models and systems
- Strong and structured way for modelling problems
- Systems turn-key solution in many cases
- Algorithmic middleware connecting smart independent components
- Base for implementing custom solutions
- Interesting research topic

Constraint programming use cases

- Combinatorial problems
 - ▶ Puzzles, combinatorial design problems, ...
- Scheduling and rostering
 - Manufacturing, Railways, Air-line plane assignments, Air-line crews, Hospital staff, . . .
- Planning
 - Vehicle routing, Laser cut path planning, Disaster evacuation, . . .
- Bioinformatics
 - Protein folding, Gene sequencing, . . .
- Testing
 - Hardware verification test planning, Covering sets, Abstract interpretation, . . .
- Various
 - Compilation, Wine blending, TV-schedule selection, Music composition, . . .

Constraint programming basics

- Define variables
- Define constraints
- Draw conclusions from constraints and current values
- When all conclusions are made
 - Make a guess
 - Draw new conclusions
 - When inconsistency detected, backtrack

Constraint programming - variables

- Finite set of variables
- Variable represents an unknown value from a set
- Finite domain variables
 - ▶ Integers E.g., x some value from $\{2, 3, 5, 7, 11, 13, 17, 19, 23, 29\}$
 - SetsE.g., y subset of {4, 8, 15, 16, 23, 42}
 - Boolean, Tasks, Graphs, Relations, Strings, . . .
- Continous domain
 - Float variables
 - Upper and lower bound, solve to certain precision

Constraint programming - constraints

- Basic constraints
 - ▶ Domain $x \in S$, $x \neq v$
 - Arithmetic $\sum_i a_i * x_i \le d$, $\sqrt{x} = y$, $x \cdot y > z$, ...
 - Element/array indexing (x[y] = z)
- Logical constraints
 - $\land ib_i = c, a \oplus b = c, \dots$
 - $x + y = z \Leftrightarrow b$ (reified constraints)
- Structural (global) constraints
 - ▶ All different $(\forall_{i,j|i\neq j}x_i \neq x_j)$
 - Global cardinality, binpacking, regular language, disjunctive and cumulative resource usage, no overlap, hamiltonian cycle, matching, minimum spanning tree, extensional . . .
 - Encapsulates re-occuring sub-structures
 - ▶ 350+ defined in Global Constraints Catalogue

Constraint programming - constraints

- Constraints are implemented as propagators
- Propagators look at current domains, and make deductions
- Example: Linear in-equality
 - $x + 2 \cdot y < z$ $x, y \in \{2..10\}, z \in \{4..10\}$
 - ▶ Propagation deduces $x \in \{2..6\}, y \in \{2..4\}, z \in \{6..10\}$
- Example: All different
 - ightharpoonup all different (x, y, z, v)
 - ▶ $x \in \{1,2\}, y \in \{1,2\}, z \in \{1,2,3,4\}, v \in \{2,4\}$
 - $x \in \{1, 2\}, y \in \{1, 2\}, z \in \{1, 2\}, z \in \{1, 2\}, v \in \{2, 4\}$
 - ► $x \in \{1, 2\}, y \in \{1, 2\}, z \in \{3, 4\}, v \in \{4\}$
 - $x \in \{1,2\}, y \in \{1,2\}, z \in \{3, X\}, v \in \{4\}$
 - $x \in \{1, 2\}, y \in \{1, 2\}, z \in \{3\}, v \in \{4\}$

Constraint programming - constraints

- Express high-level intent
- In the best case, propagation removes all variables not in any solution.
- Global constraints encapsulate smart algorithms
 - All different uses bipartite matching and strongly connected components
 - Global cardinality uses flow algorithms
 - Symmetric all different uses general matching
 - Cumulative uses edge-finding, time-tabling, and not-first/not-last reasoning
 - Binpacking uses dynamic programming
 - **.** . . .

Constraint programming - search

- Search = Branching + Exploration order
- Branching is heuristic choice
 - Defines shape of search tree
 - Smallest domain, minimum regret, smallest domain/accumulated failure count, activity based, . . .
 - Custom problem specific heuristics
- Exploration order
 - Explore search tree induced by branching
 - Depth first, Limited discrepancy, Best first, Depth bounded, . . .
 - Sequential, Restarts, Parallel, . . .
 - Large neighborhood search

Send More Money

- SEND + MORE = MONEY
- One digit per letter
- Each letter different digit
- Don't start with zeroes

SMM - Variables

```
set of int: Digits = 0..9;
var Digits: S;
var Digits: E;
var Digits: N;
var Digits: D;
var Digits: M;
var Digits: 0;
var Digits: R;
var Digits: Y;
array[1..8] of var int : letters =
  [S.E.N.D.M.O.R.Y]:
```

SMM - Constraints

```
constraint all_different(letters);
constraint 1000*S + 100*E + 10*N + D +
            1000*M + 100*O + 10*R + E =
 10000*M + 1000*0 + 100*N + 10*E + Y;
constraint S > 0:
constraint M > 0:
```

SMM - Search and output a solution

```
solve satisfy;
output [
   "S:", show(S), " E:", show(E),
   " N:", show(N), " D:", show(D),
   " M:", show(M), " 0:", show(0),
   " R:", show(R), " Y:", show(Y),
   "\n\n"
    ", show(S), show(E), show(N), show(D), "\n",
   " + ", show(M), show(O), show(R), show(E), "\n",
   " = ", show(M), show(O), show(N), show(E), show(Y),
   "\n"
];
```

SMM - Full program

```
include "globals.mzn";
set of int: Digits = 0..9:
var Digits: S:
var Digits: E:
var Digits: N;
var Digits: D;
var Digits: M;
var Digits: 0;
var Digits: R:
var Digits: Y:
array[1..8] of var int : letters =
  [S,E,N,D,M,O,R,Y];
constraint all_different(letters);
constraint 1000*S + 100*E + 10*N + D +
            1000*M + 100*O + 10*R + E =
  10000*M + 1000*O + 100*N + 10*E + Y:
constraint S > 0;
constraint M > 0;
solve satisfy:
output [
   "S:", show(S), " E:", show(E), " N:", show(N), " D:", show(D),
   " M:", show(M), " O:", show(O), " R:", show(R), " Y:", show(Y),
   "\n\n".
        ", show(S), show(E), show(N), show(D), "\n",
   " + ", show(M), show(O), show(R), show(E), "\n",
   " = ". show(M). show(O). show(N). show(E). show(Y). "\n"
];
```

SMM - Solution

```
$ mzn-gecode -a -s sendmoremoney.mzn
S:9 E:5 N:6 D:7 M:1 D:0 R:8 Y:2
    9567
   1085
 = 10652
_____
%%
   runtime: 0.000 (0.507 ms)
%%
  solvetime:
                  0.000 (0.133 \text{ ms})
%% solutions:
%% variables:
%% propagators:
%% propagations:
                  20
%%
   nodes:
%% failures:
%% restarts:
  peak depth:
```

SMM - Alternative constraints

var int: SEND; var int: MORE; var int: MONEY; constraint 1000*S + 100*E + 10*N + D = SEND;constraint 1000*M + 100*O + 10*R + E = MORE;constraint 10000*M + 1000*0 + 100*N + 10*E + Y = MONEY; constraint SEND + MORE = MONEY;

SMM - Alternative model solution

```
$ mzn-gecode -a -s sendmoremoney-alternative.mzn
S:9 E:5 N:6 D:7 M:1 D:0 R:8 Y:2
   9567
   1085
= 10652
_____
%%
   runtime: 0.012 (12.622 ms)
%% solvetime: 0.002 (2.018 ms)
%% solutions:
%% variables:
                 11
%% propagators:
%%
   propagations:
                 147
%%
   nodes:
                 11
%% failures:
%% restarts:
%% peak depth:
```

Send Most Money

- SEND + MOST = MONEY
- One digit per letter
- Each letter different digit
- Don't start with zeroes
- Maximize the amount of money

Most Money - Variables

```
set of int: Digits = 0..9;
var Digits: S;
var Digits: E;
var Digits: N;
var Digits: D;
var Digits: M;
var Digits: 0;
var Digits: T;
var Digits: Y;
array[1..8] of var int : letters =
  [S.E,N,D,M,O,T,Y];
```

Most Money - Constraints

```
constraint all different(letters);
constraint 1000*S + 100*E + 10*N + D +
            1000*M + 100*O + 10*S + T =
  10000*M + 1000*0 + 100*N + 10*E + Y;
constraint S > 0;
constraint M > 0:
var int: value =
    10000*M + 1000*0 + 100*N + 10*E + Y:
```

Most Money - Search and output

```
solve maximize value:
output [
   "S:", show(S), " E:", show(E),
   " N:", show(N), " D:", show(D),
   " M:", show(M), " 0:", show(0),
   " T:", show(T), " Y:", show(Y),
   "\n\n"
    ", show(S), show(E), show(N), show(D), "\n",
   " + ", show(M), show(O), show(S), show(T), "\n",
   " = ", show(M), show(O), show(N), show(E), show(Y),
   "\n"
];
```

Most Money - Full program

```
include "globals.mzn":
set of int: Digits = 0..9:
var Digits: S:
var Digits: E;
var Digits: N;
var Digits: D;
var Digits: M;
var Digits: 0;
var Digits: T;
var Digits: Y;
arrav[1..8] of var int : letters =
  [S.E.N.D.M.O.T.Y]:
constraint all different(letters):
constraint 1000*S + 100*E + 10*N + D +
            1000*M + 100*O + 10*R + E =
  10000*M + 1000*O + 100*N + 10*E + Y;
constraint S > 0;
constraint M > 0;
var int: value =
    10000*M + 1000*O + 100*N + 10*E + Y:
solve maximize value:
output [
   "S:", show(S), " E:", show(E), " N:", show(N), " D:", show(D),
   " M:", show(M), " 0:", show(0), " T:", show(T), " Y:", show(Y),
   "\n\n",
        ", show(S), show(E), show(N), show(D), "\n",
   " + ", show(M), show(O), show(S), show(T), "\n",
   " = ", show(M), show(O), show(N), show(E), show(Y), "\n"
1:
```

Most Money - Solution

```
$ mzn-gecode -a -s sendmostmoney.mzn
S:9 E:6 N:7 D:5 M:1 D:0 T:3 Y:8
    9675
 + 1093
 = 10768
S:9 E:7 N:8 D:3 M:1 D:0 T:2 Y:5
    9783
 + 1092
 = 10875
S:9 E:7 N:8 D:4 M:1 O:0 T:2 Y:6
    9784
 + 1092
 = 10876
```

Most Money - Statistics

```
%%
    runtime:
                    0.018 (18.043 ms)
%%
    solvetime:
                    0.003 (3.437 \text{ ms})
%% solutions:
                    8
%%
   variables:
%%
                    3
    propagators:
%%
    propagations:
                    154
%%
                    41
    nodes:
%%
    failures:
                    13
%% restarts:
%%
                    4
    peak depth:
```

Constraint programming benefits

- Concise and natural models
- Often good performance
- Custom search and hybridizations
- Describe the problem, the computer figures out how to solve it.

Constraint programming draw backs

- Complex behaviour, small changes may result in large differences
- The best model is not the simplest model
- Automatic search and symmetry breaking just starting
 - New features makes systems more complex
- Debugging constraint models is hard
- When more specialized systems work, they are often more efficient

Constraint programming alternatives

- Constraint programming as modelling language is very expressive
- Systems must handle generality
- Limiting expressiveness can lead to more effective systems
- Using CP style models translating to base language
 - Simpler models than in base language
 - Easier modelling and debugging
 - Not widely used

Linear Programming

- Restrictions
 - Only float variables
 - ▶ Only linear in-equality constraints $(\sum_i a_i * x_i \le d)$
- Mathematical optimization
- Most common method in industry and research
- Algorithms are polynomial
- Systems are very good
 - Commercial: IBM CPLEX, Gurobi, Mosek, . . .
 - ► Free: COIN-OR, lp_solve, glpk, ...
- Some things are hard to express, leading to very large models
- More information: KTH Course SF1841 Optimization

Mixed Integer Programming

- Restrictions
 - Float variables, some restricted to be integers
 - ▶ Only linear in-equality constraints $(\sum_i a_i * x_i \le d)$
- Also very common in industry and research
- Algorithms are not polynomial
- Linear relaxation (disregarding integer requirements) guides search
- Same systems as for Linear Programming
- Huge increase in performance last 15 years ($> \times 1000$)

Difference between CP and MIP models

- Consider n variables x from 1 to m and an all different constraint.
- Constraint programming

$$x = \langle x_1, x_2 \dots, x_n \rangle, x_i \in \{1, \dots, m\}$$

all different (x)

Mixed Integer Programming

$$x = \langle x_{11}, x_{12}, \dots, x_{1m}, x_{21}, \dots, x_{nm} \rangle, x_{ij} \in \{0, 1\}$$

$$\forall j \in \{1..m\} \sum_{i=1}^{n} x_{i,j} = 1 \quad \forall i \in \{1..n\} \sum_{j=1}^{m} x_{i,j} \le 1$$

SAT

- Restrictions
 - Only Boolean variables
 - Only simple or clauses
- Systems are highly optimized
- Suitable for some types of problems
- Common in industry and research
- Algorithms are not polynomial
- Explanations for failures, cheap restarts, activity based search
- More information: Print and read source of MiniSat (http://minisat.se/)

Satisfiability Modulo Theories

- Restrictions
 - Boolean variables and simple or clauses
 - Algebraic theory (equality, arithmetic, . . .)
- Makes SAT systems more usable
- Middle ground between SAT and CP
- Common in industry and research
- Algorithms are not polynomial
- Extensively used at Microsoft (https://github.com/Z3Prover/z3)

Local Search

- Move between (potentially invalid) solutions
 - Local moves and evaluation
 - Meta heuristics: Simulated annealing, Tabu search, . . .
 - ▶ Population based: genetic, ant colony, particle swarm,
- No guarantees that solution will be found
- Very often ad-hoc and unprincipled
- Can be very effective
- Constraint Based Local Search combines modelling of CP with local search
 - CBLS Systems: Comet, OscaR

Summary

- Constraint programming is a way to model and solve combinatorial optimization problems
- Even if CP systems are not used, modelling abstractions are important
- Fun way to solve puzzles
- Useful both in research and in industry
- More information: KTH Course ID2204 Constraint Programming

Bonus example - Social golfers

- Classic scheduling example
- Club with golfers playing tournament
- Each week golfers play in new groups
- All golfers play every week
- Goal: find schedule for p players in g groups of size p/g for w weeks

Social golfers example

- Golfers: Alice, Bob, Cara, David, Erica, Finley, Gretchen, Hamley, Ingrid
- Three person per group, four weeks of play
- $\{A, B, C\} \{D, E, F\} \{G, H, I\}$
- {C, D, H} {B, F, G} {A, E, I}
- $\{C, E, G\} \{B, D, I\} \{A, F, H\}$
- $\{C, F, I\}$ $\{B, E, H\}$ $\{A, D, G\}$

Set variables

- Set of integers from some universe
- $S \subseteq \{1, 2, ..., n\}$
- Simple constraints $x \in S$, |S| = y
- Set relations $S \cup T \subset U, \ldots$
- Global set constraints (all disjoint, range, roots, value precedence, partition, . . .)
- Higher level modeling

Golfers - Problem definition

Giving names to the amounts and sets

```
int: weeks;
int: groups;
int: group_size;
int: golfers = groups * group_size;
set of int: Weeks = 1..weeks;
set of int: Groups = 1..groups;
set of int: Golfers = 1..golfers;
set of int: GroupSize = 1..group_size;
```

Golfers - Data file

Stored in separate file golf443.dzn

```
% Problem instance
weeks = 4;
groups = 4;
group_size = 3;
```

Golfers - Variables

Matrix of variables representing the groups.

```
array[Weeks,Groups] of
var set of Golfers:
schedule;
```

Golfers - Right size groups

Each group must have right cardinality
constraint
forall (week in Weeks, group in Groups) (
 card(schedule[week,group]) = group_size
);

Golfers - Play once per week

Two groups in one week share no players

```
constraint
forall (week in Weeks) (
   forall (group1, group2 in Groups
           where group1 < group2) (
      schedule [week, group1]
        intersect
      schedule [week, group2]
      = {}
```

Golfers - All players every week

```
Every week all players get to play
constraint
forall (week in Weeks) (
   partition_set(
       [schedule[week,group] | group in Groups],
      Golfers
```

Golfers - No replays

Two groups from different weeks share at most one player

```
constraint
forall (week1, week2 in Weeks
        where week1 < week2) (
   forall (group1, group2 in Groups) (
      card(
         schedule[week1,group1]
            intersect
         schedule[week2,group2]
      ) <= 1
```

Golfers - Search and output

Standard search and simple output (one line per week) solve satisfy; output [if group == 1 then "Week " ++ show(week) ++ ": " else "" endif ++ show(schedule[week,group]) ++ " " ++ if group == groups then "\n" else "" endif | week in Weeks, group in Groups 1;

Golfers - First solution

```
$ mzn-gecode -s social.mzn golf443.dzn
Week 1: {5,11,12} {3,4,8} {2,7,10} {1.6.9}
Week 2: {6,7,11} {3,5,10} {2,4,9} {1,8,12}
Week 3: {6,10,12} {3,9,11} {2,5,8} {1,4,7}
Week 4: {8,10,11} {7,9,12} 4..6 1..3
%%
   runtime:
                 29.833 (29833.078 ms)
%% solvetime:
                  29.820 (29820.188 ms)
%% solutions:
%% variables:
                  248
%% propagators:
               220
%%
                 74940759
   propagations:
   nodes:
                  762337
%% failures:
                  381153
%% restarts:
                  0
%%
                  41
   peak depth:
```

Golfers - Parallel search

```
$ mzn-gecode -s -p 4 social.mzn golf443.dzn
Week 1: {5,7,12} {3,4,9} {2,6,10} {1,8,11}
Week 2: {5,9,10} {3,11,12} {2,4,8} {1,6,7}
Week 3: {6,9,12} {3,7,8} {2,5,11} {1,4,10}
Week 4: {8,10,12} {7,9,11} 4..6 1..3
%%
   runtime: 1.138 (1138.283 ms)
%% solvetime: 1.126 (1126.825 ms)
%% solutions:
%% variables:
                  248
%% propagators:
               220
%%
                  4959569
   propagations:
   nodes:
                  52228
%% failures:
                  26077
%% restarts:
                  0
%%
   peak depth:
                  39
```

Symmetry breaking

- Social golfers has symmetries
 - Solutions can be transformed simply
 - Only care about some solution
- Golfer names symmetric
- Order of groups in week is symmetric
- Order of weeks is symmetric

Golfer names symmetric

- Base solution
 - ► {*A*, *B*, *C*} {*D*, *E*, *F*} {*G*, *H*, *I*}
 - $\{C, D, H\} \{B, F, G\} \{A, E, I\}$
 - $\{C, E, G\} \{B, D, I\} \{A, F, H\}$
 - $ightharpoonup \{C, F, I\} \{B, E, H\} \{A, D, G\}$
- Switched A and F
 - $\{F, B, C\} \{D, E, A\} \{G, H, I\}$
 - $\{C, D, H\} \{B, A, G\} \{F, E, I\}$
 - $\{C, E, G\} \{B, D, I\} \{F, A, H\}$
 - $ightharpoonup \{C, A, I\} \{B, E, H\} \{F, D, G\}$

Group order symmetry

- Base solution
 - ► {A, B, C} {D, E, F} {G, H, I}
 - $\{C, D, H\} \{B, F, G\} \{A, E, I\}$
 - $\{C, E, G\} \{B, D, I\} \{A, F, H\}$
 - $\{C, F, I\} \{B, E, H\} \{A, D, G\}$
- Re-ordered first two groups
 - $\{D, E, F\} \{A, B, C\} \{G, H, I\}$
 - ► {C, D, H} {B, F, G} {A, E, I}
 - ► {*C*, *E*, *G*} {*B*, *D*, *I*} {*A*, *F*, *H*}
 - $\{C, F, I\} \{B, E, H\} \{A, D, G\}$

Week order symmetry

- Base solution
 - ► {A, B, C} {D, E, F} {G, H, I}
 - $\{C, D, H\} \{B, F, G\} \{A, E, I\}$
 - $\{C, E, G\} \{B, D, I\} \{A, F, H\}$
 - $\{C, F, I\} \{B, E, H\} \{A, D, G\}$
- Re-ordered first and last week
 - $\{A, B, C\} \{D, E, F\} \{G, H, I\}$
 - ► {C, D, H} {B, F, G} {A, E, I}
 - $\{C, E, G\} \{B, D, I\} \{A, F, H\}$
 - $\{C, F, I\} \{B, E, H\} \{A, D, G\}$

Golfers - Group order symmetry

Idea: Groups in week can be ordered constraint forall (week in Weeks) (forall (group1, group2 in Groups where group1 < group2) (schedule[week, group1] > schedule[week, group2]

Golfers - Group order symmetry

```
$ mzn-gecode -s social.mzn golf443.dzn
Week 1: {1,7,11} {2,5,10} {3,6,9} {4,8,12}
Week 2: {1,6,10} {2,4,9} {3,7,8} {5,11,12}
Week 3: {1,4,5} {2,6,7} {3,10,12} {8,9,11}
Week 4: 1..3 {4,10,11} {5,6,8} {7,9,12}
%%
   runtime:
                 1.090 (1090.510 ms)
%%
   solvetime:
                  1.086 (1086.382 ms)
%% solutions:
%% variables:
                  248
%% propagators:
               244
%%
                  2639492
   propagations:
                  24363
   nodes:
%%
   failures:
                  12172
   restarts:
                  0
%%
   peak depth:
                  38
```

Golfers - First week symmetry

```
Idea: First week can be fixed statically

constraint
forall (group in Groups, i in GroupSize) (
          ((group-1)*group_size + i)
                in
                schedule[1,group]
);
```

Golfers - First week symmetry

```
$ mzn-gecode -s social.mzn golf443.dzn
Week 1: 1...3 4...6 7...9 10...12
Week 2: {6,9,10} {3,4,8} {2,7,11} {1,5,12}
Week 3: {6,7,12} {3,5,11} {2,4,9} {1,8,10}
Week 4: {6,8,11} {3,9,12} {2,5,10} {1,4,7}
%%
   runtime: 0.177 (177.424 ms)
%% solvetime:
                 0.166 (166.398 ms)
%% solutions:
%% variables:
              248
%% propagators: 213
%%
                  412250
   propagations:
                  4071
   nodes:
%% failures:
                 2025
%% restarts:
                  0
%%
                  27
   peak depth:
```

Golfers - Search annotation

```
Idea: Search on first group per week first
solve ::
   set search(
      [schedule[week,1] | week in Weeks] ++
      [schedule[week,group]
       | week in Weeks, group in Groups],
      input_order, indomain_min, complete)
satisfy;
```

Golfers - Search annotation

```
$ mzn-gecode -s -p 4 social.mzn golf443.dzn
Week 1: {1,10,11} {2,5,7} {3,6,12} {4,8,9}
Week 2: {1,6,7} {2,4,12} {3,8,11} {5,9,10}
Week 3: {1,4,5} {2,6,11} {3,7,9} {8,10,12}
Week 4: 1..3 {4,7,10} {5,6,8} {9,11,12}
%%
   runtime:
                 0.036 (36.828 ms)
%% solvetime:
                  0.030 (30.782 ms)
%% solutions:
%% variables:
                  248
%% propagators: 244
%%
                 101909
   propagations:
                  1030
   nodes:
%% failures:
                  493
%% restarts:
                  0
%%
   peak depth:
                  35
```

Golfers - All features

```
$ mzn-gecode -s social.mzn golf443.dzn
Week 1: 1...3 4...6 7...9 10...12
Week 2: {1,4,7} {2,5,10} {3,9,11} {6,8,12}
Week 3: {1,5,8} {2,6,11} {3,7,12} {4,9,10}
Week 4: {1,6,9} {2,4,12} {3,8,10} {5,7,11}
%%
   runtime:
                 0.016 (16.187 ms)
%% solvetime:
                  0.004 (4.421 \text{ ms})
%% solutions:
%% variables: 248
%% propagators:
                231
%%
                  4625
   propagations:
                  63
   nodes:
%% failures:
                  21
%% restarts:
                  0
%% peak depth:
                  23
```

Golfers - All features and parallel

```
$ mzn-gecode -s -p 4 social.mzn golf443.dzn
Week 1: 1..3 4..6 7..9 10..12
Week 2: {1,4,7} {2,5,10} {3,9,11} {6,8,12}
Week 3: {1,5,8} {2,6,11} {3,7,12} {4,9,10}
Week 4: {1,6,9} {2,4,12} {3,8,10} {5,7,11}
%%
   runtime: 0.008 (8.116 ms)
%% solvetime:
                  0.002 (2.617 \text{ ms})
%% solutions:
%% variables: 248
%% propagators:
                231
%%
                  4693
   propagations:
   nodes:
                  64
%% failures:
                  22
%% restarts:
                  0
%%
   peak depth:
                  23
```

Social golfers conclusions

- High level modeling with set variables
- Suprisingly hard problem
- Parallel search cheap to test
- Symmetry breaking is crucial but hard
 - Symmetry breakings might interfere with each other
 - Wrong order for group order symmetry would make first week fixing invalid
- Search is an art
 - Experimentation often required