

# Facets of multiple *all\_different* predicates

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We explore the properties of the comb structure of multiple all different predicates by presenting two classes of facet-inducing inequalities of the convex hull of integer solutions. We present experimental results based on these inequalities in a hybrid Constraint Programming - Integer Programming solver.

- ▶ Timetabling Example
- ▶ The `alldifferent` Constraint
- ▶ Mathematical Formulation
- ▶ New Facets
- ▶ Separation Algorithm
- ▶ Experimental results

# A timetabling problem

- ▶ Students choose a number of courses
- ▶ Each course may have many sections
- ▶ Instructors teach a number of courses
- ▶ Each section meets sometime
- ▶ Each section meets in some room

Assign students, instructors, room to meeting times

# NP-hard if ...

- ▶ ... only one student
- ▶ ... only one section per course
- ▶ ... only one instructor per course
- ▶ ... only one meeting per section
- ▶ ... only one room per section

And multiple combinations of the above

## In practice, even worse

- ▶ Must satisfy student requests
- ▶ Students are busy 6 periods per day
- ▶ Students must have lunch period
- ▶ Some course must be paired
- ▶ Some courses must avoid certain periods
- ▶ Meetings must be spread out during week
- ▶ Instructors have limits on meetings per day
- ▶ Instructors have limits on adjacent meetings

Of course, there is no feasible solution!

# Elements of a Model

## Data

- ▶ John: MTH105, MTH204, MTH218
- ▶ Mary: MTH105, MTH206, MTH210, MTH219
- ▶ Prof. Cheng: MTH206, MTH218

## Model

- ▶  $x_i$  contains timeslots of courses of  $i$
- ▶ `alldifferent`{ $x_{John}$ }
- ▶ `alldifferent`{ $x_{Mary}$ }
- ▶ `alldifferent`{ $y_{Cheng}$ }
- ▶ Moreover, meeting assignments must be coherent

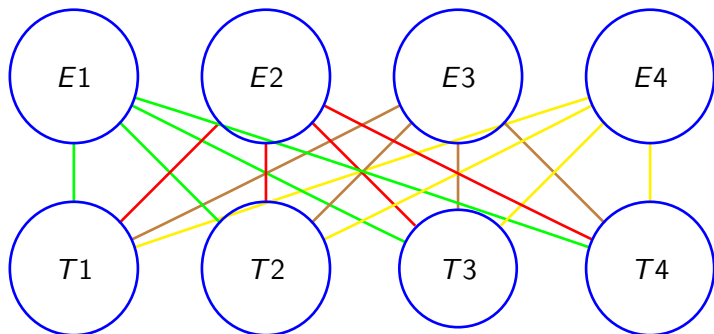
`alldifferent` is the workhorse of such a model

## Definition

The **alldifferent constraint** on the collection of variables  $x_1, x_2, \dots, x_n$  with respective finite domains  $D_1, D_2, \dots, D_n$  is defined as:

$$\begin{aligned} & \text{alldifferent}\{x_1, \dots, x_n\} \\ & = \\ & \{(d_1, \dots, d_n) \mid d_i \in D_i, i \neq j \Rightarrow d_i \neq d_j\} \end{aligned}$$

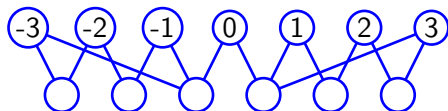
# Graph of *one* alldifferent



# Algorithms for one alldifferent

- ▶ Bounds consistency
- ▶ Range consistency
- ▶ Hyper-arc consistency

# HAC of alldifferent is not enough



$\{\{0,1\},\{1,2\},\{2,3\},\{0,3\},\{0,-1\},\{-1,-2\},\{-2,-3\},\{0,-3\}\}$

Simultaneous Matching problem

Let us restrict ourselves to combs

# CP Formulation of Comb Structure

- ▶  $D = \{0, 1, \dots, k\} \subset \mathbb{Z}$
- ▶  $x_i, x_i^j \in D$

`alldifferent`{ $x_1, x_2, \dots, x_n$ }

`alldifferent`{ $x_1, x_1^j : j \in T_1$ }

`alldifferent`{ $x_2, x_2^j : j \in T_2$ }

⋮

`alldifferent`{ $x_n, x_n^j : j \in T_n$ }

Handle  
Tooth 1  
Tooth 2  
⋮  
Tooth  $n$

# IP Formulation of Comb Structure

- ▶  $P_y$  is polytope of the simultaneous matchings of the comb structure of multiple all different:

$$x_i = \sum_r r y_{ir} \quad \forall i \in H$$

$$x_i^j = \sum_r r y_{ir}^j \quad \forall j \in T_i$$

$$\sum_r y_{ir} = 1 \quad \forall i \in H$$

$$\sum_i y_{ir} \leq 1 \quad \forall r \in D$$

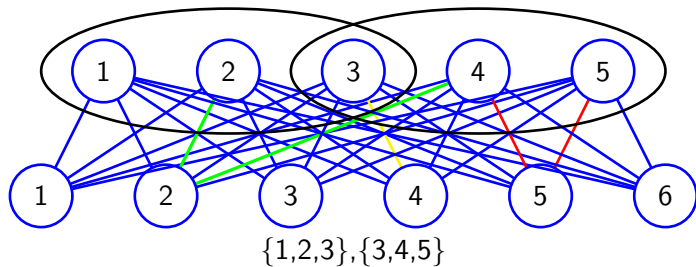
$$\sum_r y_{ir}^j = 1 \quad \forall j \in T_i$$

$$\sum_i y_{ir}^j \leq 1 \quad \forall r \in D$$

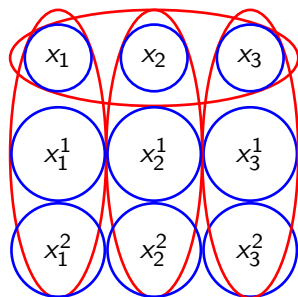
$$y_{ir}, y_{ir}^j \in \{0, 1\} \quad \forall i \in H, \quad \forall j \in T_i, \quad \forall r \in D$$

- ▶ The integral polytope  $P_I$  is the projection of  $P_y$  onto the  $x$ -space

# As a graph



# Comb Structure of Timetabling Example



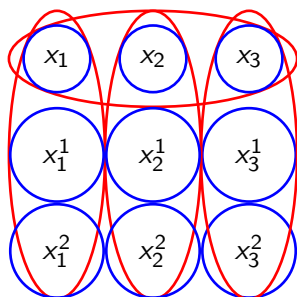
$\text{alldifferent}\{x_1, x_2, x_3\}$   
 $\text{alldifferent}\{x_1, x_1^1, x_1^2\}$   
 $\text{alldifferent}\{x_2, x_2^1, x_2^2\}$   
 $\text{alldifferent}\{x_3, x_3^1, x_3^2\}$

Handle  
Tooth 1  
Tooth 2  
Tooth 3

New predicate?

$\text{alldifferent}\{\{x_1, x_2, x_3\}, \{x_1, x_1^1, x_1^2\}, \{x_2, x_2^1, x_2^2\}, \{x_3, x_3^1, x_3^2\}\}$

# Timetabling Example Solutions



Some solutions include:

$$\begin{array}{cccc} (x_1, x_2, x_3, & x_1^1, x_1^2, & x_2^1, x_2^2, & x_3^1, x_3^2) \\ = (1, 2, 3, & 4, 5, & 4, 5, & 4, 5) \end{array}$$

and

$$\begin{array}{cccc} (x_1, x_2, x_3, & x_1^1, x_1^2, & x_2^1, x_2^2, & x_3^1, x_3^2) \\ = (2, 5, 3, & 0, 1, & 0, 4, & 0, 1) \end{array}$$

Our facet-inducing inequalities for the multiple alldifferent comb structure extend the work done by previous researchers including:

- ▶ H.P. Williams and H. Yan who provided facets for the 1-alldifferent constraint in their paper titled “Representations of the alldifferent Predicate of Constraint Satisfaction in Integer Programming”
- ▶ G. Appa, D. Magos, and I. Mourtos who provided facets for the 2-alldifferent constraint in their paper titled “On the system of two alldifferent predicates”

# New Facets of $P_I$

The following are facet-inducing inequalities for the comb structure whenever we exceed the minimum number of domain values.

$$\sum_{i \in S} (x_i + |S| \sum_{j \in S_i} x_i^j) \geq \sum_{m=|S_i|}^{|S_i|+|S|-1} m + |S|^2 \sum_{m=0}^{|S_i|-1} m,$$
$$\forall S \subseteq H : |S| > 1, \forall S_i \subseteq T_i : i \in S, |S_i| = s$$

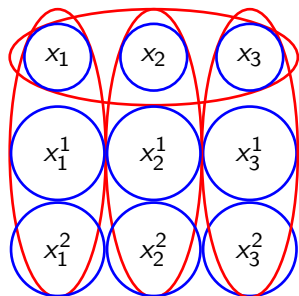
(1)

$$\sum_{i \in S} (x_i + |S| \sum_{j \in S_i} x_i^j) \leq \sum_{m=k-|S_i|-|S|+1}^{k-|S_i|} m + |S|^2 \sum_{m=k-|S_i|+1}^k m$$
$$\forall S \subseteq H : |S| > 1, \forall S_i \subseteq T_i : i \in S, |S_i| = s$$

(2)

# Facets of Timetabling Example

- ▶  $D = \{0, 1, \dots, 5\}$
- ▶  $LB \leq \sum_{i \in S} (x_i + |S| \sum_{j \in S_i} x_i^j) \leq UB$

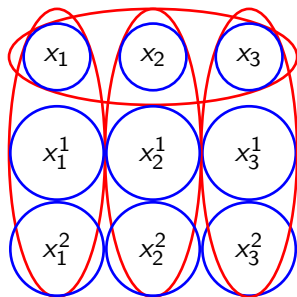


- ▶  $S = \{x_1, x_3\}, S_1 = \{x_1^1\}, S_3 = \{x_3^2\}$ 
  - ▶  $3 \leq x_1 + x_3 + 2x_1^1 + 2x_3^2 \leq 27$
- ▶  $S = \{x_1, x_3\}, S_1 = \{x_1^1, x_1^2\}, S_3 = \{x_3^1, x_3^2\}$ 
  - ▶  $9 \leq x_1 + x_3 + 2x_1^1 + 2x_1^2 + 2x_3^1 + 2x_3^2 \leq 41$

# Outline of Proof of Facets of $P_I$

To show that inequalities (1) and (2) are facet-inducing, we prove two things:

- ▶ They are valid inequalities
  - ▶ Assign domain values appropriately to show that the lower and upper bounds are satisfied.



$$9 \leq x_1 + x_3 + 2x_1^1 + 2x_1^2 + 2x_3^1 + 2x_3^2 \leq 41$$

# Outline of Proof of Facets of $P_I$

- ▶ Next, we show that any other valid inequalities can be obtained by taking nonnegative linear combinations of (1) and (2).
  - ▶ We define the set  $F_1$  where:

$$F_1 = \left\{ x \in P_I : \sum_{i \in S} (x_i + |S| \sum_{j \in S_i} x_i^j) = \sum_{m=|S_i|}^{|S_i|+|S|-1} m + |S|^2 \sum_{m=0}^{|S_i|-1} m \right\}$$

- ▶ Let  $A$  denote the set indexing all the variables in the system.

# New Facets Proof

## Theorem

For  $a \in \mathfrak{R}^{|A|}$ ,  $a_0 \in \mathfrak{R}$ , if  $ay = a_0$  holds  $\forall y \in F_1$ , there exists a scalar  $\pi$  such that:

$$a_i = \begin{cases} \pi, & i \in S \\ 0, & i \in H - S \end{cases} \quad (3)$$

$$a_i^j = \begin{cases} |S|\pi, & j \in S_i \\ 0, & j \in T_i - S_i \end{cases} \quad (4)$$

and

$$a_0 = \pi \left[ \sum_{m=|S_i|}^{|S_i|+|S|-1} m + |S|^2 \sum_{m=0}^{|S_i|-1} m \right]. \quad (5)$$

# Separation Algorithm

$\pi = \text{sort}(\bar{x}_1, \dots, \bar{x}_n)$ ,  $\pi_i = \text{sort}(\bar{x}_i^1, \dots, \bar{x}_i^{s_i})$

**for**  $i = 2 \dots n$  **do**

$L = \sum_{l=n-i+1}^n \bar{x}(\pi(l))$ ;  $U = \sum_{l=1}^i \bar{x}(\pi(l))$

**for**  $j = 1 \dots \min\{s_1, \dots, s_n\}$  **do**

**for**  $m = 1, \dots, i$  **do**

$L = L + i * \bar{x}_{\pi(n-m+1)}^{\pi_{\pi(n-m+1)}(s_{\pi(n-m+1)}-j+1)}$ ;  $U = U + i * \bar{x}_{\pi(m)}^{\pi_{\pi(m)}(j)}$

**end for**

**if**  $L < \sum_{m=j}^{j+i-1} m + i^2 \sum_{m=0}^{j-1} m =: L_{ij}$  **then**

$\sum_{l=n-i+1}^n x(\pi(l)) + i \sum_{m=1}^i x_{\pi(n-m+1)}^{\pi_{\pi(n-m+1)}(s_{\pi(n-m+1)}-j+1)} \geq L_{ij}$

**end if**

**if**  $U > \sum_{m=k-j-i+1}^{k-j} m + i^2 \sum_{m=k-j+1}^k m =: U_{ij}$  **then**

$\sum_{l=1}^i x(\pi(l)) + i \sum_{m=1}^i x_{\pi(m)}^{\pi_{\pi(m)}(j)} \leq U_{ij}$

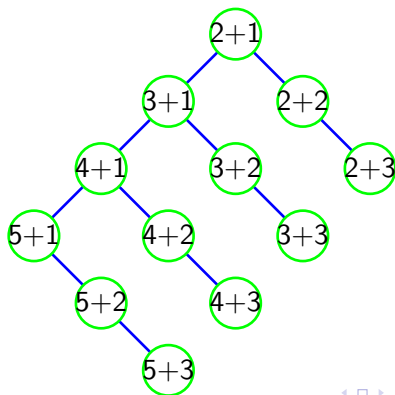
**end if**

**end for**

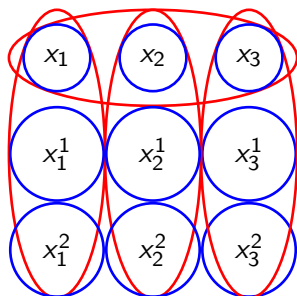
**end for**

# Separation Algorithm

- ▶ Add variables from the handle one by one
- ▶ Add corresponding teeth variable one by one
- ▶ Compare value to bound
- ▶ Complexity:  $|H| * \max_i |T_i|$
- ▶ Our algorithm implicitly constructs the tree:



# Applying our Separation Algorithm to Timetabling Example



- ▶ Optimizing  $x_1 + x_2 + x_3 + 3x_1^2 + 3x_2^2 + 3x_3^2 + 3x_1^1 + 3x_2^1 + 3x_3^1$  over the all different facet-inducing inequalities known up to now gives an objective function value of 88.5 and the point:

$$\begin{aligned} & (x_1, x_2, x_3, x_1^1, x_2^1, x_3^1, x_1^2, x_2^2, x_3^2, x_1^1, x_2^1, x_3^1) \\ & = (2.5, 2.5, 2.5, 4, 5, 4, 5, 5, 4) \end{aligned}$$

- ▶ Not a solution

# How the Separation Algorithm Works

Obtained point:

$$\begin{array}{ccccccccc} (x_1, & x_2, & x_3, & x_1^1, & x_1^2, & x_2^1, & x_2^2, & x_3^1, & x_3^2) \\ = (2.5, & 2.5, & 2.5, & 4, & 5, & 4, & 5, & 5, & 4) \end{array}$$

- ▶ Sort the variables in descending order within (Handle, Tooth 1, Tooth 2, Tooth 3)

$$\begin{array}{ccccccccc} (x_1, & x_2, & x_3, & x_1^2, & x_1^1, & x_2^2, & x_2^1, & x_3^1, & x_3^2) \\ = (2.5, & 2.5, & 2.5, & 5, & 4, & 5, & 4, & 5, & 4) \end{array}$$

# How the Separation Algorithm Works

- ▶ Check for violated inequalities in this order:

- ▶ **Facet 2 + 1:**  $x_1 + x_2 + 2x_1^2 + 2x_2^2 = 25 \leq 27$

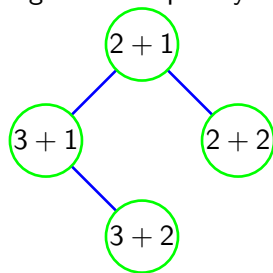
- ▶ **Facet 2 + 2:**  $x_1 + x_2 + 2x_1^2 + 2x_2^2 + 2x_1^1 + 2x_2^1 = 41 \leq 41$

- ▶ **Facet 3 + 1:**  $x_1 + x_2 + x_3 + 3x_1^2 + 3x_2^2 + 3x_3^1 = 52.5 \leq 54$

- ▶ **Facet 3 + 2 :**

$$x_1 + x_2 + x_3 + 3x_1^2 + 3x_2^2 + 3x_3^1 + 3x_1^1 + 3x_2^1 + 3x_3^2 = 88.5 \not\leq 87$$

The following tree is implicitly constructed:



# Are these any good?

Experimental setup:

- ▶ CP solver : ECLiPSe
- ▶ LP solver : Coin-OR Clp

TK-YW	YW
1.000	0.921
1.000	0.984
0.991	0.951
1.000	0.982
0.995	0.970
0.992	0.982
0.984	0.967
0.972	0.963
0.955	0.945
0.938	0.931

Table: Ratio of LP relaxation and best CP solution after 10 seconds

# Random Timetabling Instances

Experimental data pending ...

- ▶ More general facets
- ▶ Separation algorithm based on Network Flow subproblem
- ▶ Variations on alldifferent

# Open questions (partial answers)

- ▶ New predicate `alldifferent`  $\{H, T_0, T_1, \dots, T_n\}$ ?
  - ▶ Not for CP formulations; maybe for IP
- ▶ Do better than bounding the objective?
  - ▶ Yes, we can use dual values to prune search tree
- ▶ Recognize 'good' combs amongst multiple `alldifferent`?
  - ▶ Maybe use the cuts on different structures