

# Integer and Constraint Programming Revisited for Mutually Orthogonal Latin Squares (Student Abstract)

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## Abstract

We use integer programming (IP) and constraint programming (CP) to search for sets of mutually orthogonal latin squares (MOLS). We build upon the work of Appa *et al.* in their paper "Searching for Mutually Orthogonal Latin Squares via Integer and Constraint Programming"[1] by formulating an extended symmetry breaking method and providing an alternative CP encoding which performs much better in practice.

## Orthogonal Latin Squares

A *latin square* of order  $n$  is an  $n \times n$  array,  $L$ , of symbols  $\{0, 1, \dots, n-1\}$  in which each symbol appears exactly once in each row and column. The entry in row  $i$  and column  $j$  of a square  $L$  is denoted  $L_{ij}$ . Two latin squares of the same order,  $L$  and  $M$ , are said to be *orthogonal* if there is a unique solution  $L_{ij} = a$ ,  $M_{ij} = b$  for every pair of  $a, b \in \{0, 1, \dots, n-1\}$ . A set of  $k$  latin squares of order  $n$ , is called a set of *mutually orthogonal latin squares* (MOLS) if all squares are pairwise orthogonal—in which case we label the system as  $k\text{MOLS}(n)$ .

0	1	2	3
1	0	3	2
2	3	0	1
3	2	1	0

0	1	2	3
2	3	0	1
3	2	1	0
1	0	3	2

0	1	2	3
3	2	1	0
1	0	3	2
2	3	0	1

Figure 1: An example of 3MOLS(4).

## Symmetry Breaking

Solutions to the  $k\text{MOLS}(n)$  problem are isomorphic to many other solutions, so we impose constraints on the domains of cells to reduce symmetries. Let  $X, Y$  be a set of  $2\text{MOLS}(n)$ , we impose the following:

1. Fix first row of  $X$  and  $Y$  in lex order to prevent permutations of symbols within squares.
2. Fix first column of  $X$  in lex order to prevent permutations of rows across all squares.
3. Fix first column of  $Y$  to be one of  $a_n$  tuples with distinct cycle types.

The last strategy follows from Theorem 1 in our supplementary submission material, where a detailed proof is also provided. The value  $a_n$  is given in entry A002865 of the Online Encyclopedia of Integer Sequences[2] and has growth rate  $e^{O(\sqrt{n})}$ .

## References

- [1] G. Appa, D. Magos, and I. Mourtos. Searching for mutually orthogonal latin squares via integer and constraint programming. *European Journal of Operational Research*, 173(2):519–530, September 2006.
- [2] OEIS Foundation Inc. *The On-Line Encyclopedia of Integer Sequences*, 2021. <http://oeis.org/A002865>.
- [3] Charles F. Laywine and Gary L. Mullen. *Discrete mathematics using Latin squares*. John Wiley & Sons, 1998.
- [4] Laurent Perron and Vincent Furnon. OR-Tools, 2019. <https://developers.google.com/optimization/>.
- [5] Gurobi Optimization, LLC. Gurobi Optimizer Reference Manual. <https://www.gurobi.com/documentation/9.1/refman/>.

## IP Model

Our IP model for  $2\text{MOLS}(n)$  uses  $n^4$  binary variables

$$x_{ijkl} := \begin{cases} 1 & \text{if } X_{ij} = k \text{ and } Y_{ij} = l \\ 0 & \text{otherwise} \end{cases}$$

for all  $i, j, k, l \in \{0, 1, \dots, n-1\}$ . The Latin and orthogonality constraints are expressed as

$$\begin{aligned} \sum_{0 \leq k, l < n} x_{ijkl} &= 1 \quad \forall i, j && 1 \text{ value per cell} \\ \sum_{0 \leq j, l < n} x_{ijkl} &= 1 \quad \forall i, k && \text{Latin rows of } X \\ \sum_{0 \leq j, k < n} x_{ijkl} &= 1 \quad \forall i, l && \text{Latin rows of } Y \\ \sum_{0 \leq i, l < n} x_{ijkl} &= 1 \quad \forall j, k && \text{Latin columns of } X \\ \sum_{0 \leq i, k < n} x_{ijkl} &= 1 \quad \forall j, l && \text{Latin columns of } Y \\ \sum_{0 \leq i, j < n} x_{ijkl} &= 1 \quad \forall k, l && X, Y \text{ orthogonal} \end{aligned}$$

## CP Model

Our CP model for  $2\text{MOLS}(n)$  uses  $2n^2$  integer variables

$X_{ij} :=$  value of cell  $(i, j)$  in square  $X$ ,  
 $Y_{ij} :=$  value of cell  $(i, j)$  in square  $Y$ ,  
 $Z_{ij} :=$  value of cell  $(i, j)$  in square  $Z = X^{-1}Y$

Where  $i, j, X_{ij}, Y_{ij}, Z_{ij} \in \{0, 1, \dots, n-1\}$  and the  $i$ th row of  $Z$  is row  $i$  of  $Y$  applied to the inverse of row  $i$  of  $X$ .  $X, Y$  are orthogonal if and only if  $Z$  is Latin [3, Theorem 6.6], so we impose orthogonality by ensuring  $X, Y$  and  $Z$  are Latin squares.

AllDifferent( $X_{ij} \forall j$ ), AllDifferent( $X_{ij} \forall i$ ),  
 AllDifferent( $Y_{ij} \forall j$ ), AllDifferent( $Y_{ij} \forall i$ ),  
 AllDifferent( $Z_{ij} \forall j$ ), AllDifferent( $Z_{ij} \forall i$ )

To make  $Y = XZ$  the  $(i, j)$ th entry of  $Y$  is set to the  $(i, X_{ij})$ th entry of  $Z$  using “element indexing” constraints  $Z_i[X_{ij}] = Y_{ij}$  where  $Z_i$  is the vector of variables corresponding to row  $i$  of  $Z$ .

## Results

We ran trials on a computer with an Intel i9 9900k processor and 32GB of memory. Trials were allocated 1 core each and timeout was set at 60,000s. The implementations of our programs were done in Microsoft Visual C++ and later modified to run in Linux. We used Gurobi[5] as our IP solver and Google OR-Tools[4] as our CP solver. Each are highly competitive in their class, and are free to use for students. The CP-linear model given by Appa *et al.* [1] was originally used, which imposed orthogonality by defining  $Z_{ij} := X_{ij} + nY_{ij}$  and AllDifferent( $Z_{ij} \forall i, j$ ). We later revised this to our CP-index model, which outperformed all other models even with no symmetry breaking.

Model	5	6	7	8	9	10
IP	0.1	Timeout	3.2	6.4	344.5	3,046.4
CP-linear	0.0	Timeout	8.0	1,967.1	58,637.8	Timeout
CP-index	0.0	Timeout	7.8	36.3	378.7	214.8

Table 1: Timings in seconds for orders  $5 \leq n \leq 10$  with no symmetry breaking.

Imposing all of the symmetry breaking strategies gave us significant improvements in the running time of all models. Symmetry breaking strategy 3 is an extension of the strategy used by Appa *et al.* [1], who show that any solution to  $k\text{MOLS}(n)$  is isomorphic to one where  $Y_{10} = 2$ ,  $Y_{i0} \neq i$  and  $Y_{i0} \leq i + 1$  for  $1 \leq i < n$ , with the total number of fixings given by the  $(n-2)^{\text{th}}$  Fibonacci number. Imposing symmetry breaking strategy 3 reduces this number to  $a_n$ .

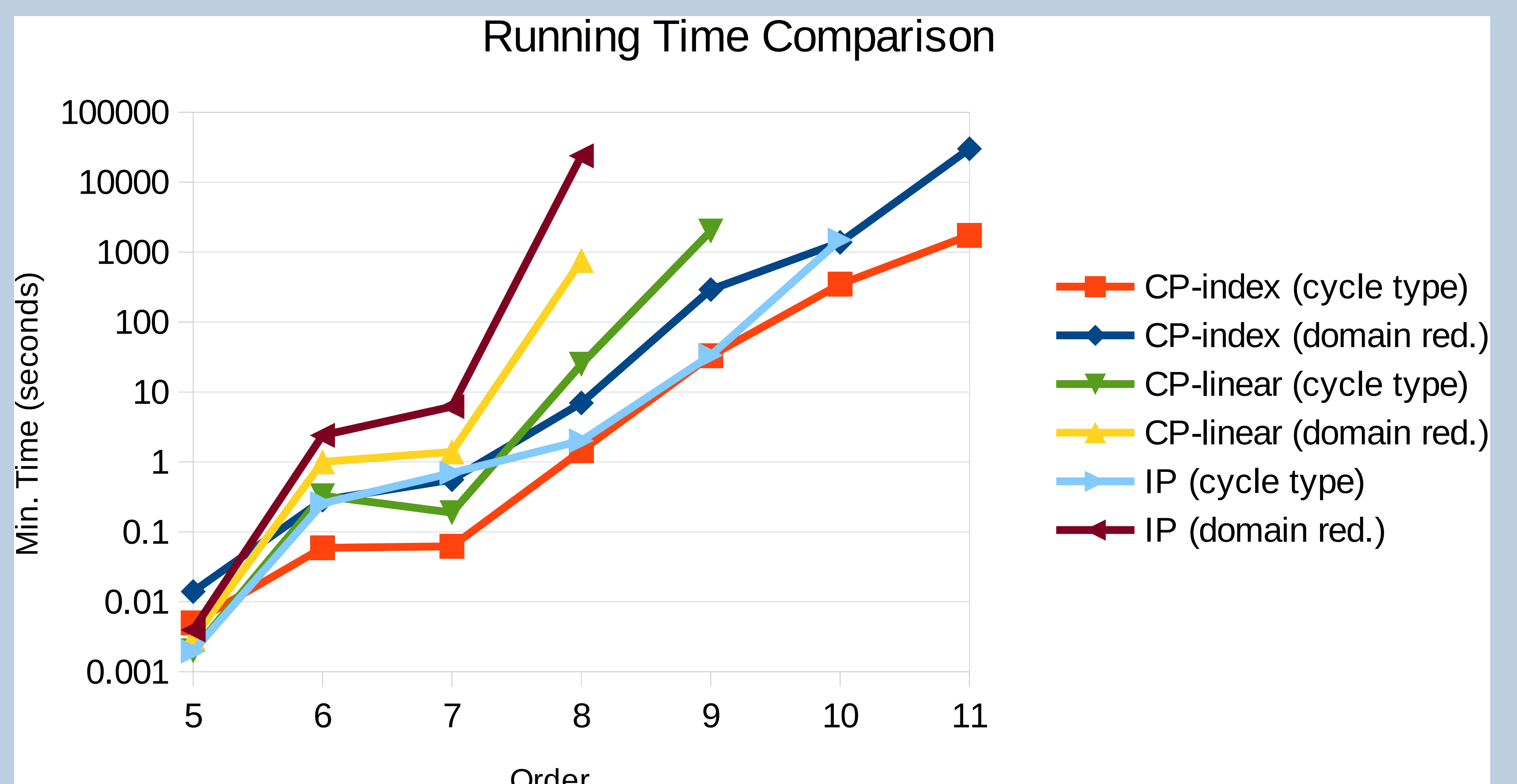


Figure 2: Running times of our models and symmetry breaking methods for  $5 \leq n \leq 11$ .

More detailed timings, proofs and complete implementations of our programs can be found at <https://github.com/noahrubin333/CP-IP>.