



**AUTOMATICALLY IMPROVING CONSTRAINT  
MODELS IN SAVILE ROW THROUGH  
ASSOCIATIVE-COMMUTATIVE COMMON  
SUBEXPRESSION ELIMINATION**

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# MOTIVATION – THE LONG-TERM GOAL

Automated Reformulation – Given a naive model, automatically improve it via a sequence of reformulations

We can take inspiration from two main sources:

- Compilers
- Automating techniques used by expert modellers

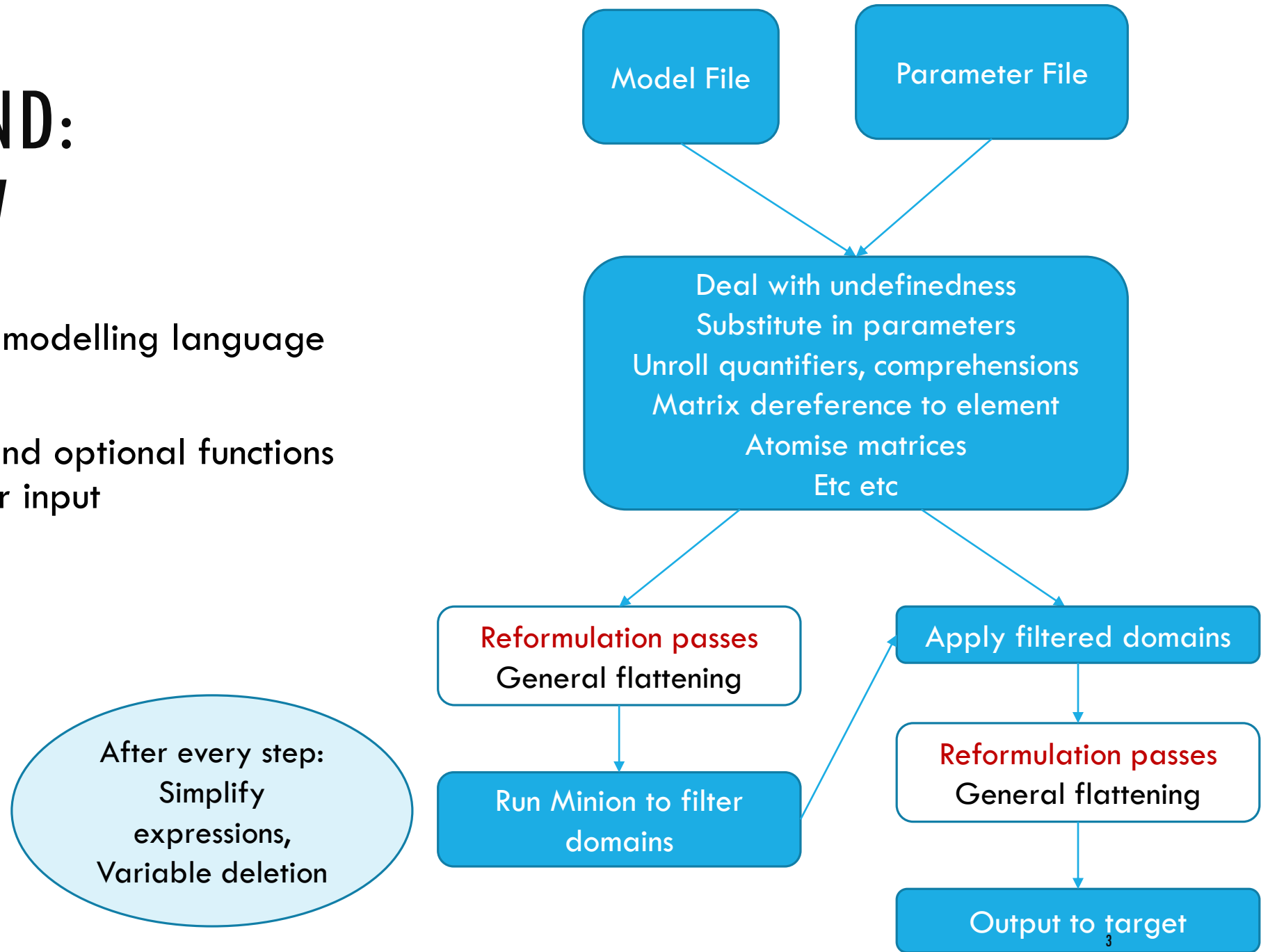
Typically stronger reformulations can be done on problem instances

This paper describes an algorithm for an instance-level reformulation – AC-CSE

# BACKGROUND: SAVILE ROW

A system that takes modelling language  
'Essence'

Performs essential and optional functions  
to translate to solver input



# BACKGROUND: IDENTICAL COMMON SUB-EXPRESSION ELIMINATION

The most basic **Common Sub-Expression Elimination (CSE)** extracts **identical expressions**

$$x+y \leq 4, \quad x+y=z$$

$$\text{where } x,y \text{ in } \{0..5\}, z \text{ in } \{0..10\}$$

Extract  $x+y$  and replace with new *auxiliary variable*

$$x+y=aux, \quad aux \leq 4, \quad aux=z$$

**This allows the constraint solver to see that  $z \leq 4$**

Without CSE, the solver can discover that  $z \leq 8$  (via  $x \leq 4$  and  $y \leq 4$ )

CSE connects together overlapping constraints via a new variable

# ASSOCIATIVE-COMMUTATIVE COMMON SUBEXPRESSION ELIMINATION

Associative-Commutative Common Subexpression Elimination (AC-CSE) uses the fact that sum, product, conjunction, disjunction are **associative and commutative**

In practice treat sums, products, conjunctions and disjunctions as unordered sets

Extract common subsets from sets of expressions

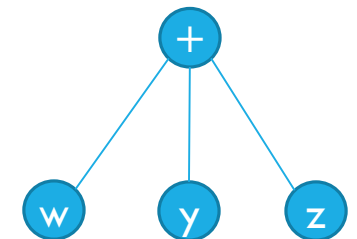
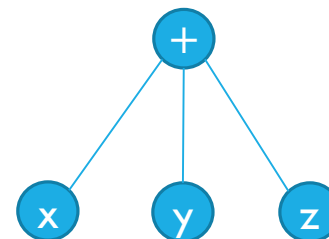
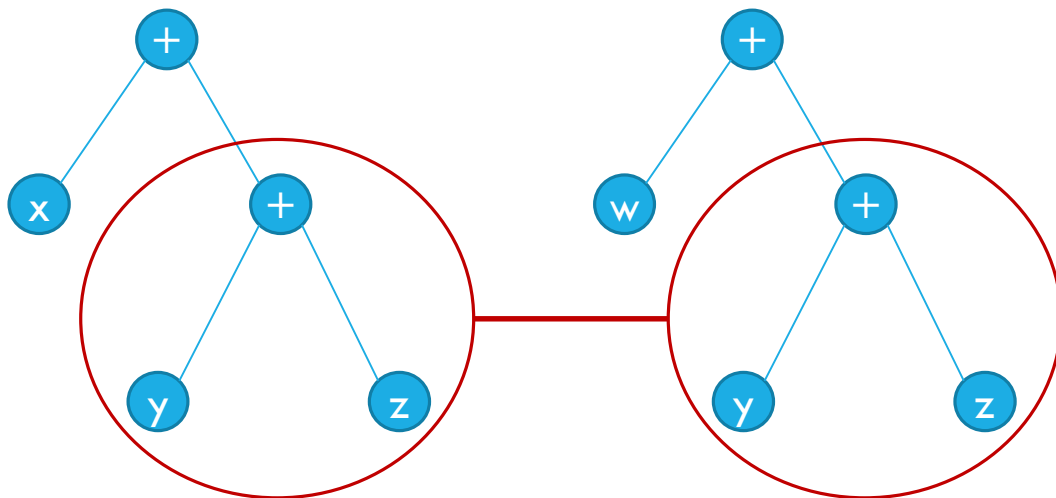
# ASSOCIATIVE-COMMUTATIVE CSE

We have already sorted associative-commutative expressions (normalisation)

- $x+y+z$  matches  $z+x+y$  with Identical CSE

But we cannot yet match arbitrary overlaps

- A binary tree representation would allow matching prefixes (left-branching) or postfixes (right-branching) in the sorted order... GNU C++ compiler does this
- ...but Savile Row represents AC operators using non-binary trees

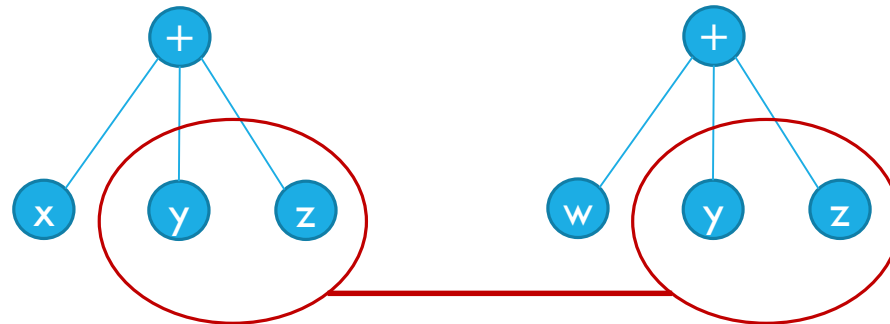


# ASSOCIATIVE-COMMUTATIVE CSE

Treat an AC expression as a **set of terms**

Find a subset common to two or more AC expressions

Extract the common subset everywhere and replace with an auxiliary variable



Can improve propagation dramatically

With some sensible assumptions, never reduces propagation

# ASSOCIATIVE-COMMUTATIVE CSE

Extracting one AC-CS may block others

X-CSE (our proposed algorithm) uses heuristic ordering

- Extracts AC-CS with most occurrences first
- Never copies original expressions – can be more efficient in finite-domain context

I-CSE [Araya et al, CP 2008] extracts **all** AC-CSs between two expressions

- Makes copies of original expressions – potential big slowdown
- Context: Numerical CSP

Genuine choice – difficult to know right answer

Our experiments suggest X-CSE better in finite-domain context



# EXPERIMENTS

Comparing total time (solving plus Savile Row time)

With AC-CSE (implemented by X-CSE algorithm) vs same config without AC-CSE

Many other options switched on: domain filtering, variable deletion, etc

Minion as the solver

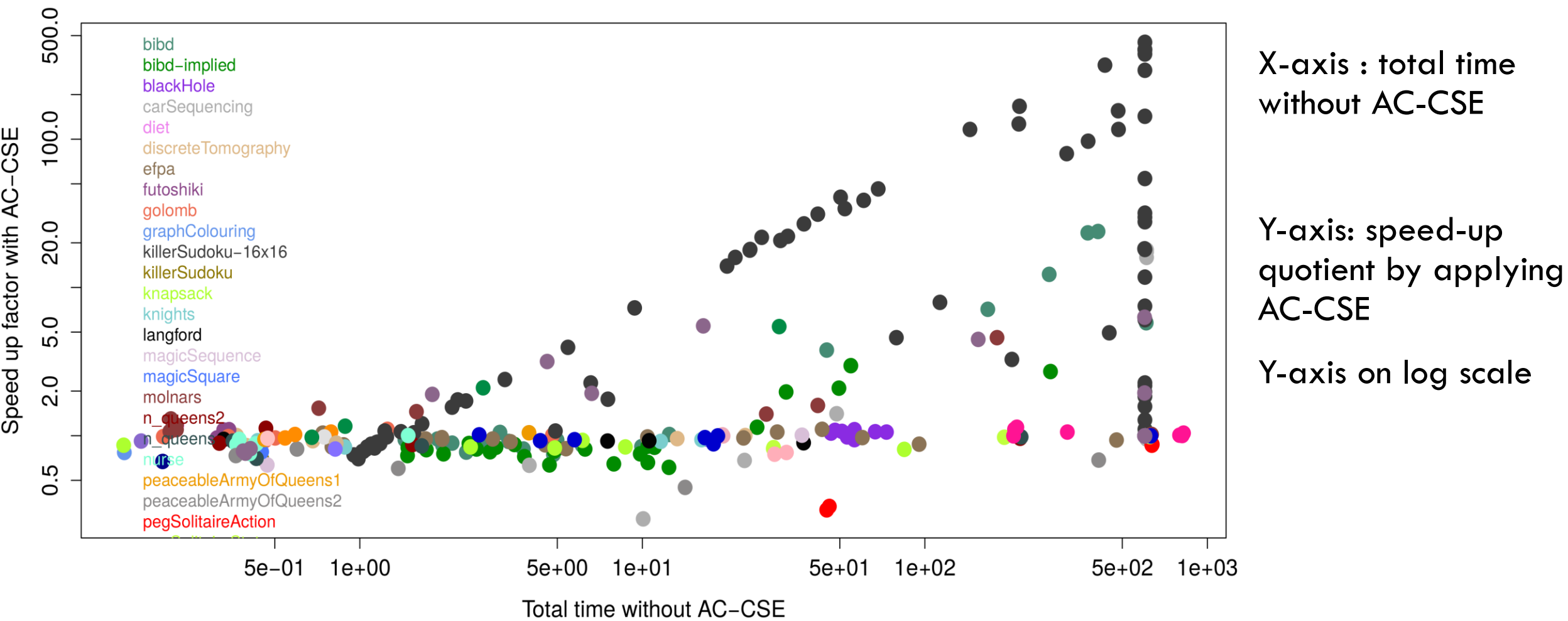
- Static variable, value orderings

**AC-CSE never increases search**

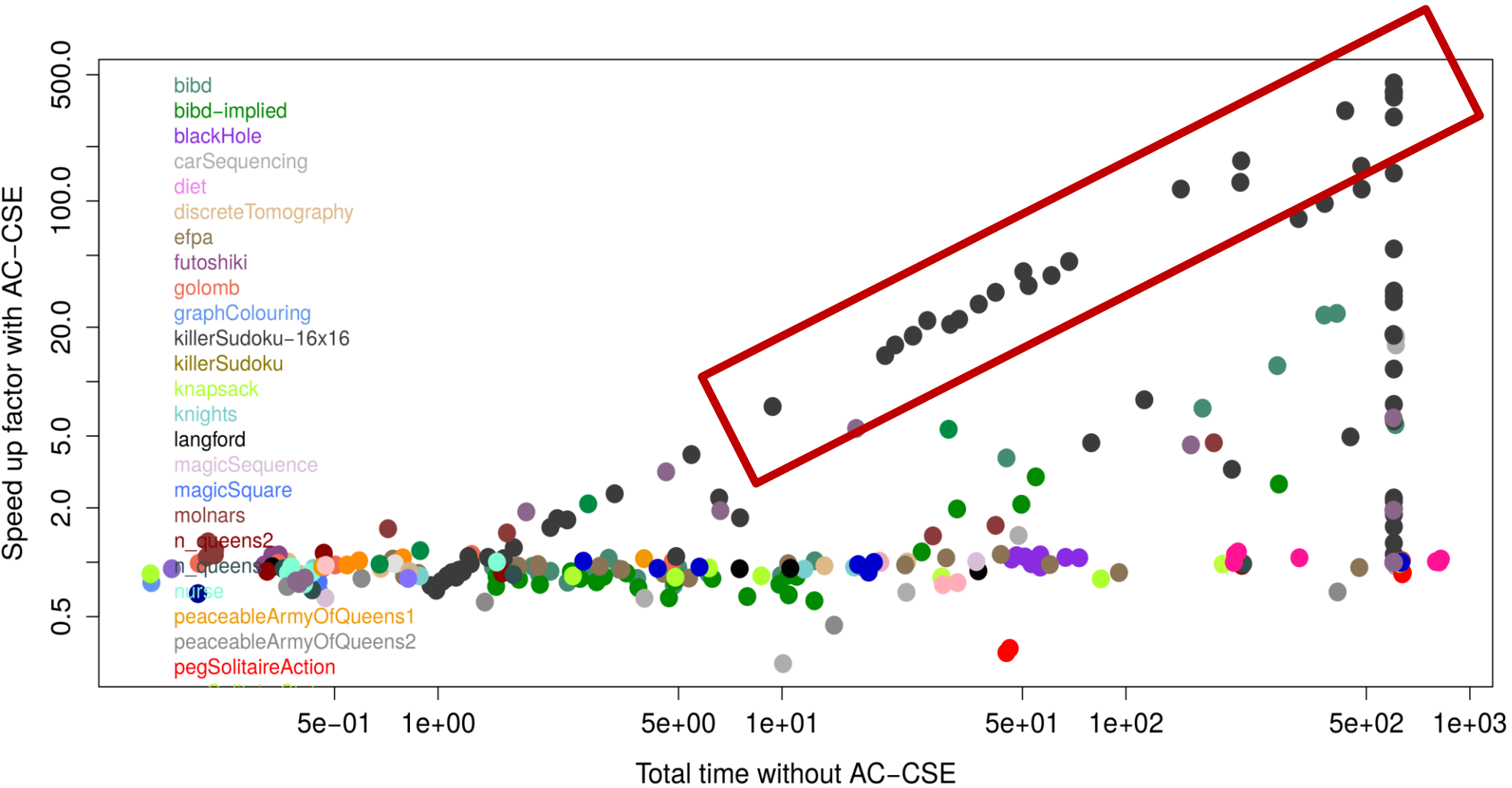
- **But can increase Minion time, Savile Row time.**

Results on following slides are from a slightly newer Savile Row than paper

# EXPERIMENTAL RESULTS



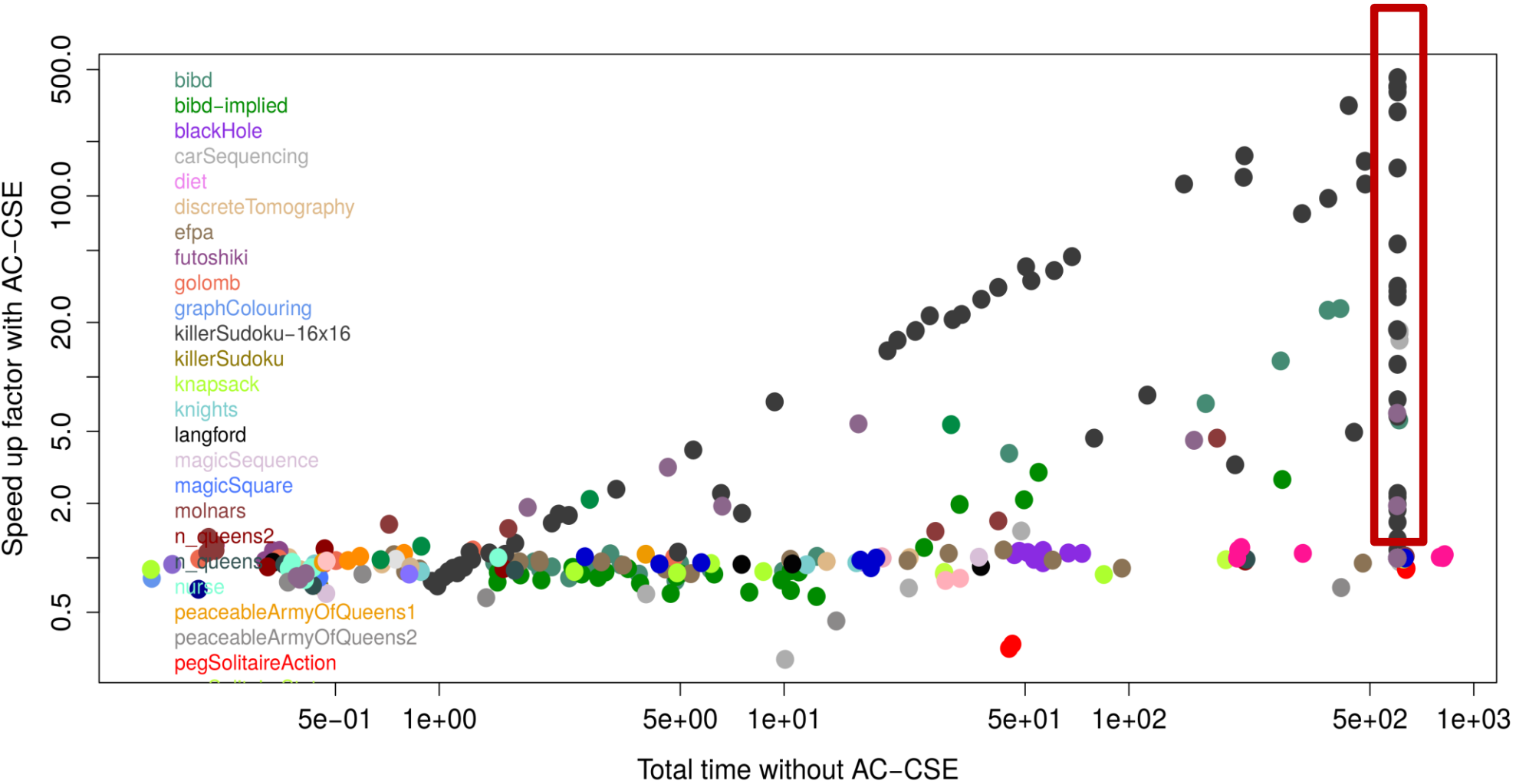
# EXPERIMENTAL RESULTS



Some Killer Sudoku instances are rendered almost trivial by AC-CSE

Seems to have exponential growth in speed-up quotient

# EXPERIMENTAL RESULTS



Some instances solve within 10 minutes with AC-CSE, time-out without it

Killer Sudoku (24)

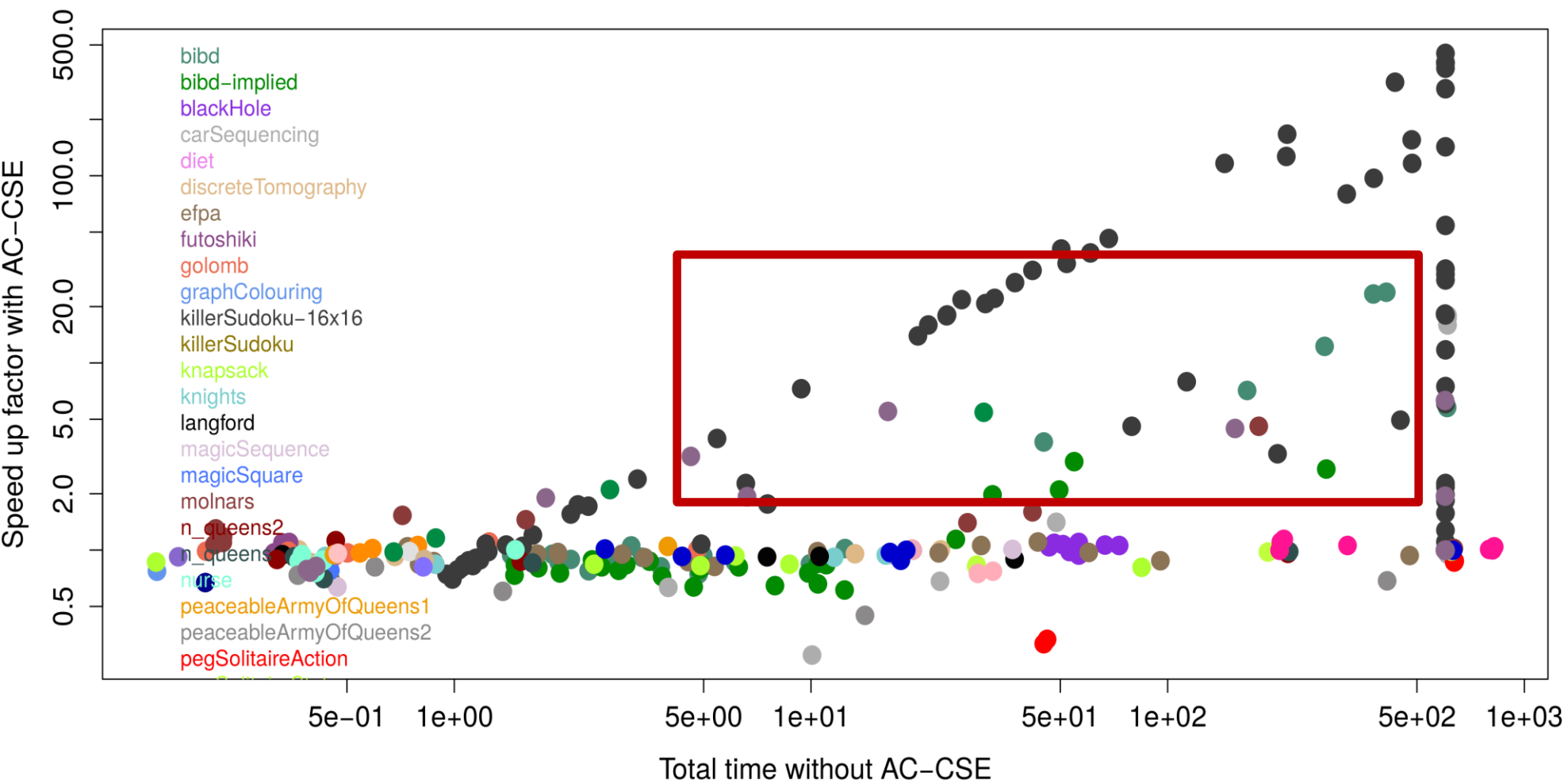
Car Sequencing (3)

SONET (2)

BIBD (2)

BIBD-implied (1)

# EXPERIMENTAL RESULTS



The rest above 2x

BIBD (5)

BIBD-implied (3)

Killer Sudoku (5)

SONET (3)

Molnars (1)

Waterbucket (2)



# KILLER SUDOKU

9x9 grids too easy, we did 16x16

16x16 matrix, each cell takes value in {1..16}

Rows, columns and 4x4 subsquares: allDifferent

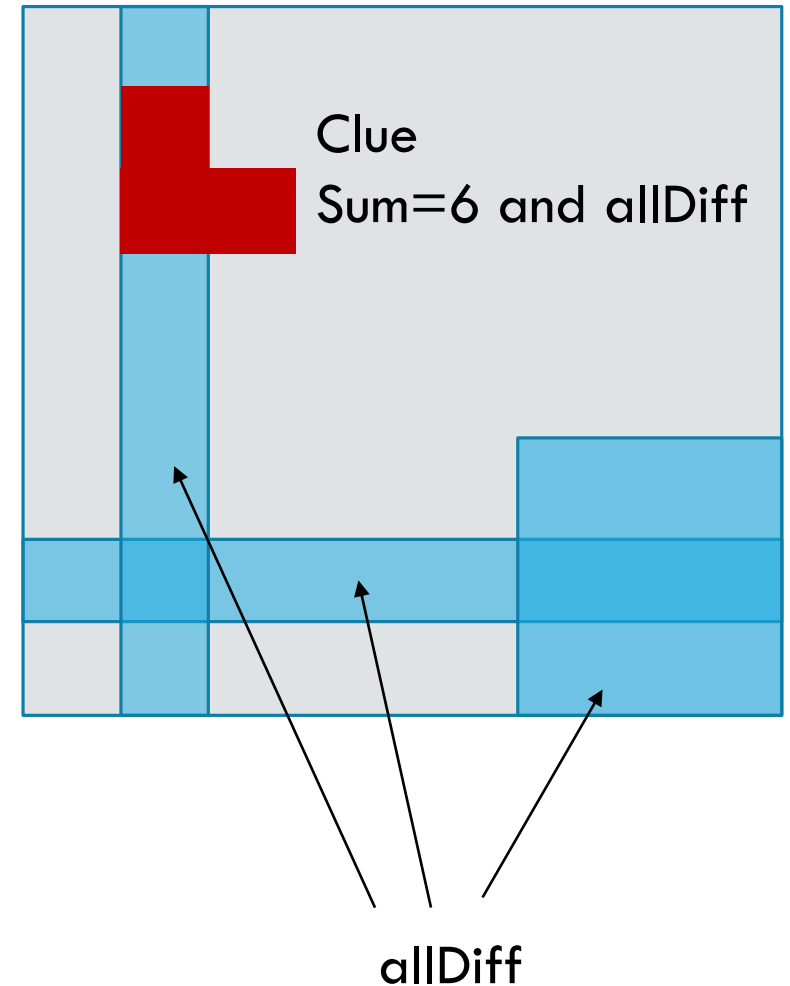
Clues are contiguous sets of cells

- The sum is given as part of the clue
- Cells within the clue are allDifferent

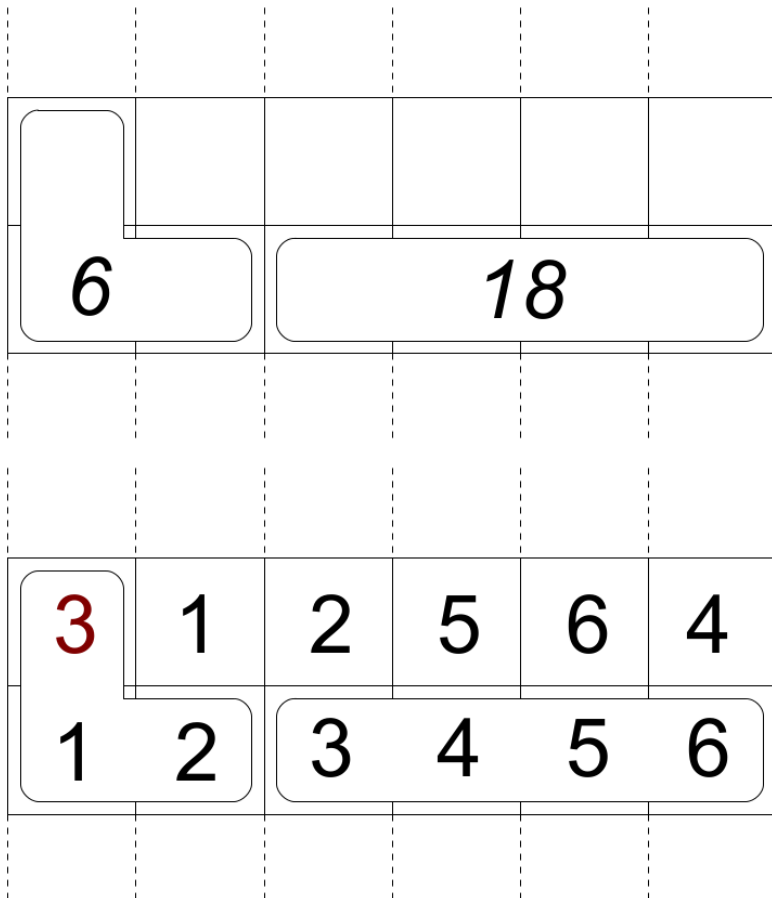
In the example (right) the clue must contain values 1,2,3

Entire matrix covered by non-overlapping clues

Model is exactly the above constraints



# KILLER SUDOKU



Rows/columns/subsquares are a permutation

**Introduce sum constraints from AllDifferent**

For each row, column and subsquare  $X$ :

$$\text{sum}(X) = 136 \quad (\text{for } 16 \times 16 \text{ case})$$

Suppose we had 6x6 Killer Sudoku (left)

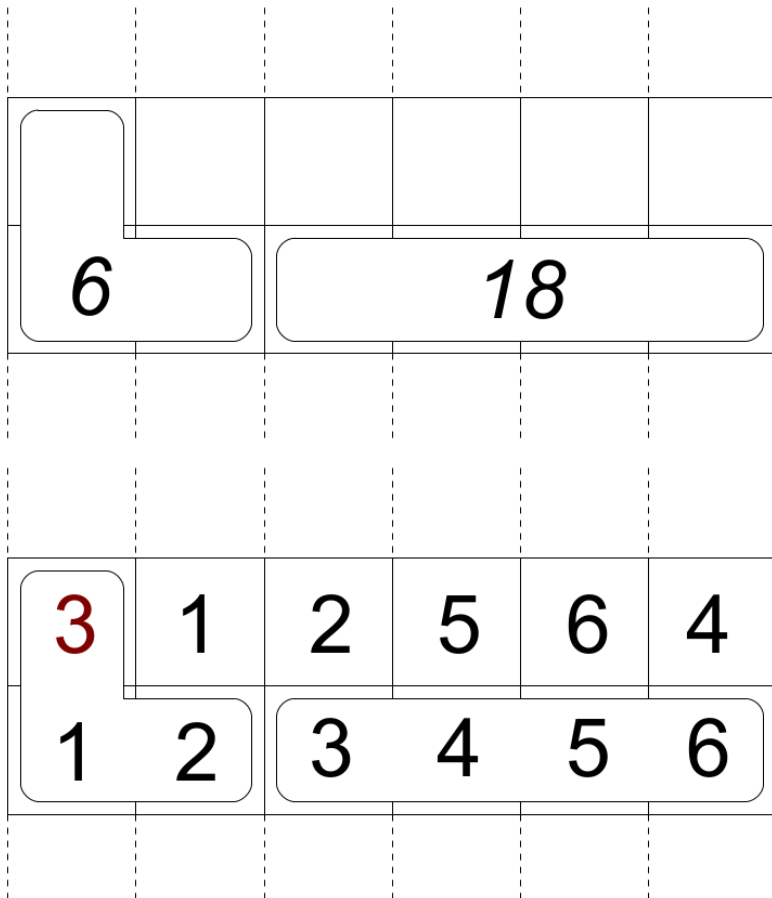
$$\text{sum}(X) = 21$$

For each clue, we also get useless  $\text{sum} \leq a$  and  $\text{sum} \geq b$

- Removed by Identical CSE followed by simplifiers



# KILLER SUDOKU



New sums on rows/columns/subsquares intersect with clues

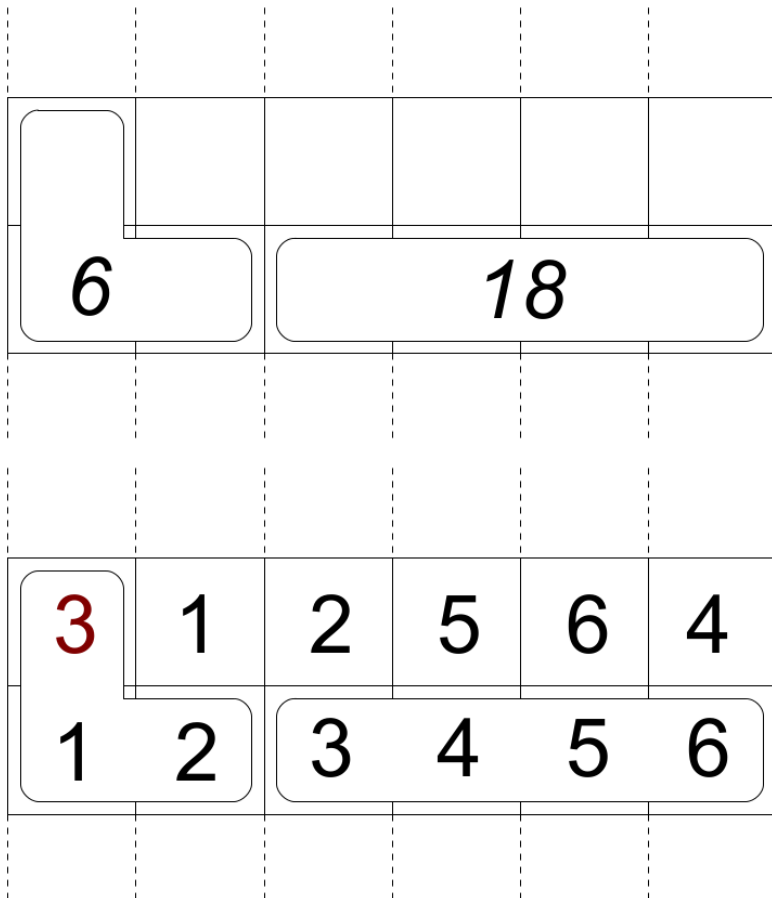
In example (left), suppose two rows are  $k[1,..]$  and  $k[2,..]$

**AC-CSE connects clues to rows**

$k[2,3] + \dots + k[2,6]$  is common to the 18 clue and the row sum

$k[2,1] + k[2,2]$  is common to the 6 clue and the row sum

# KILLER SUDOKU



$$k[2,3] + \dots + k[2,6] = \text{aux1}$$

$$k[2,1] + k[2,2] = \text{aux2}$$

$$\text{aux1} = 18, \text{aux2} + k[1,1] = 6$$

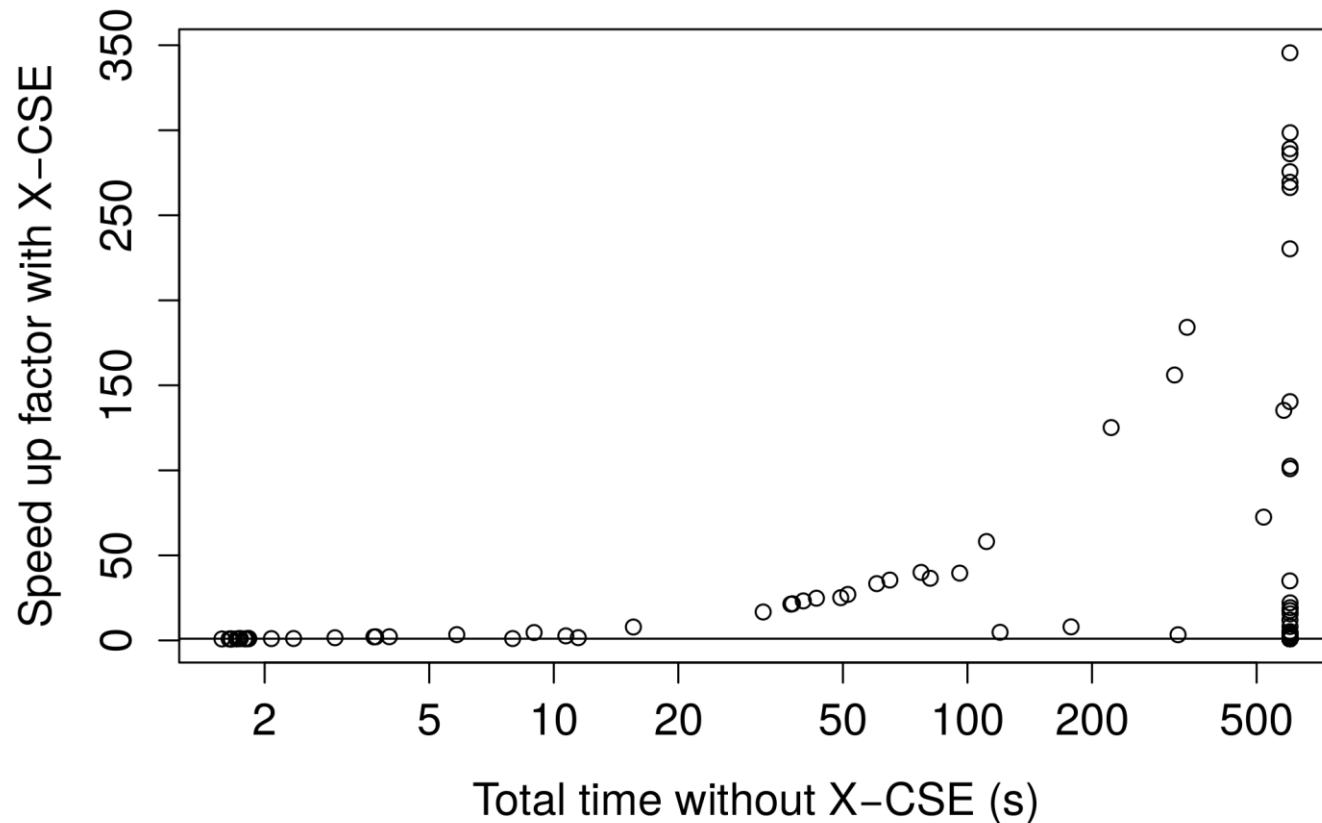
$$\text{aux1} + \text{aux2} = 21$$

**aux1** replaced with 18 (variable deletion)

**aux2** becomes 3 (simplifier, then var deletion)

$k[1,1]$  becomes 3 (simplifier, then var deletion)

# KILLER SUDOKU – RESULTS



Some hard problems made almost trivial

Peak instance:

**Without X-CSE** Savile Row took 2.26s

Minion timed out at 600s

2,774,028 nodes

**With X-CSE** Savile Row took 1.62s

Minion took 0.13s, 2 nodes

`savilerow -O3 killer.eprime ...`

# KILLER SUDOKU – SUMMARY

We need to do these steps:

1. Add implied sum to all AllDifferent constraints
2. Apply AC-CSE
3. Variable deletion (interleaved with simplifiers)

# FUTURE WORK

## Scalability

Both X-CSE and I-CSE can fail if...

- Very large number of sum constraints
- Very long sums

Many possible ways of improving scalability

- Anytime algorithm with time quota

Obtain more implied sum (or product) constraints from globals

# THANK YOU

AC-CSE is implemented in Savile Row:

<http://savilerow.cs.st-andrews.ac.uk/>

Any questions?