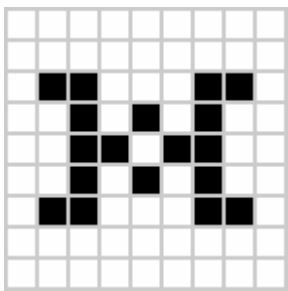
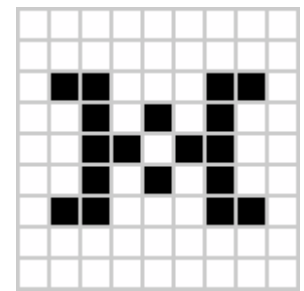


Generating Special-purpose Stateless Propagators for Arbitrary Constraints

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Why is this the way forward?

- **59,049**
- **564,092,290**

How do you propagate an arbitrary new constraint?

- Hope it is one built in to your solver?
 - But it may not be
- Encode it using existing constraints?
 - Requires modelling time
 - May not achieve GAC
- Use table constraint implementation?
 - Can be exponential during search
 - Definitely not optimised to new constraint
- Write a new special purpose propagator?
 - At best takes programmer time
 - At worst is a significant research project

Instead, we show how to ...

- Generate a special purpose propagator for an arbitrary new constraint
 - Either specified by table or existing propagator
- The generated propagator is ...
 - created by code-generating code
 - guaranteed to achieve GAC every call
 - worst case time of $O(nd)$ per call during search
- Our best experimental results
 - Better than handwritten propagators
 - Achieved 18x speedup on “Oscillating Life”
 - Carrying out identical search to table constraint
 - Easily repaid overheads many times over

But ...

- The overheads can be ...
 - Exponential preprocessing before search
 - Exponential space needed during search
 - 10-ary boolean constraint close to our limit
- Overheads incurred per different table
 - So need constraint to occur many times
 - In a single problem
 - Or over many problems
 - Which does happen
 - Oscillating Life, compile constraint once, run often

DOMINION

- This work is part of the Dominion project
 - Generously funded by EPSRC
 - <http://dominion.cs.st-andrews.ac.uk>
 - Follow up to Minion
- *Generates* solver for every instance
 - i.e. is code-generating code
 - Produces a very special purpose solver!
 - Exploit individual features of instance
 - Make implementation choices for this instance
 - Which may not be good in a general purpose solver

DOMINION

- For generating GAC propagators ...
 - We therefore know domain size of all variables
 - And exact set of allowed tuples
 - *But* we will often see that propagators are reusable
- For technical reasons ...
 - Propagators used in Minion, not Dominion
 - Technical reason being Dominion not existing

Basic Idea

- For *every* state domains of variables can be in
 - Compute what GAC does and store it
 - Either from propagator we are given ...
 - ... or from GAC-table propagator
- Generate code which
 - Looks up state of domains
 - Then does what the store tells it to

Basic Idea

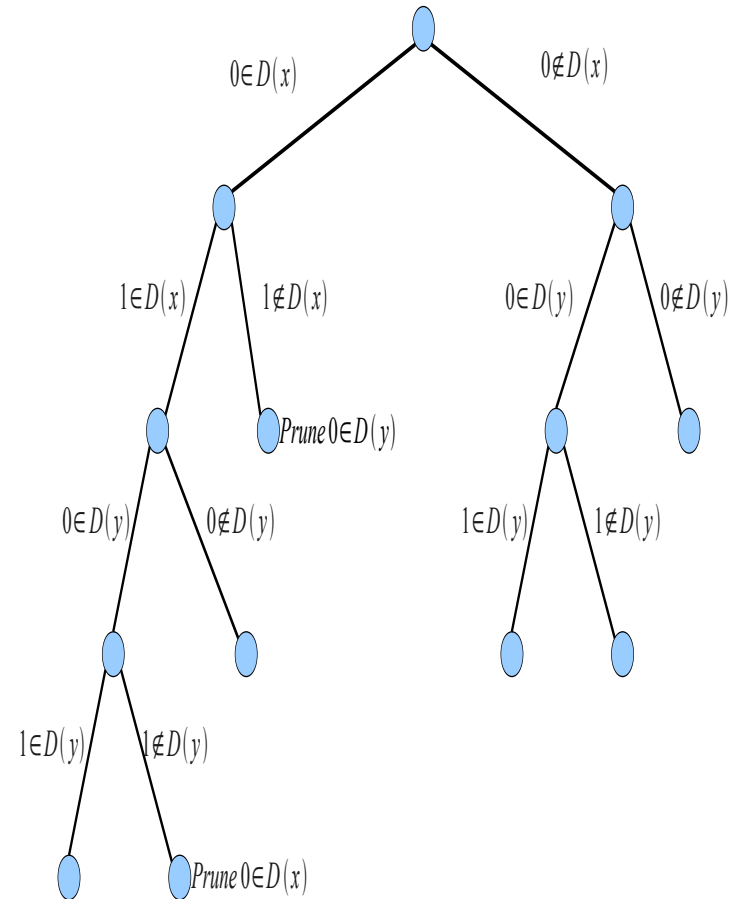
- Constraint of arity n , domain size d
- There are $2^d - 1$ states of each variable
 - Excluding the empty domain
- So there are $(2^d - 1)^n$ states of all variables
 - e.g. $d = 2, n = 10$ gives 59,049 states
- So it will pay to avoid considering all states
- Note that algorithm is *stateless*
 - i.e. it stores no state between invocations
 - All variants in this paper have this property

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Second idea

- Build a binary decision tree
- Nodes store propagations
 - And next decision (unless leaf)
 - Leaves are sure to be GAC
- Decisions are domain membership
 - i in $D(x)$? for variable x value i
 - E.g. if domain of x is $\{1,3\}$
 - Then 1 in $D(x)$, 2 not in $D(x)$, 3 in $D(x)$
 - Even though 1 or 3 may later be removed
- Proved correct in this paper



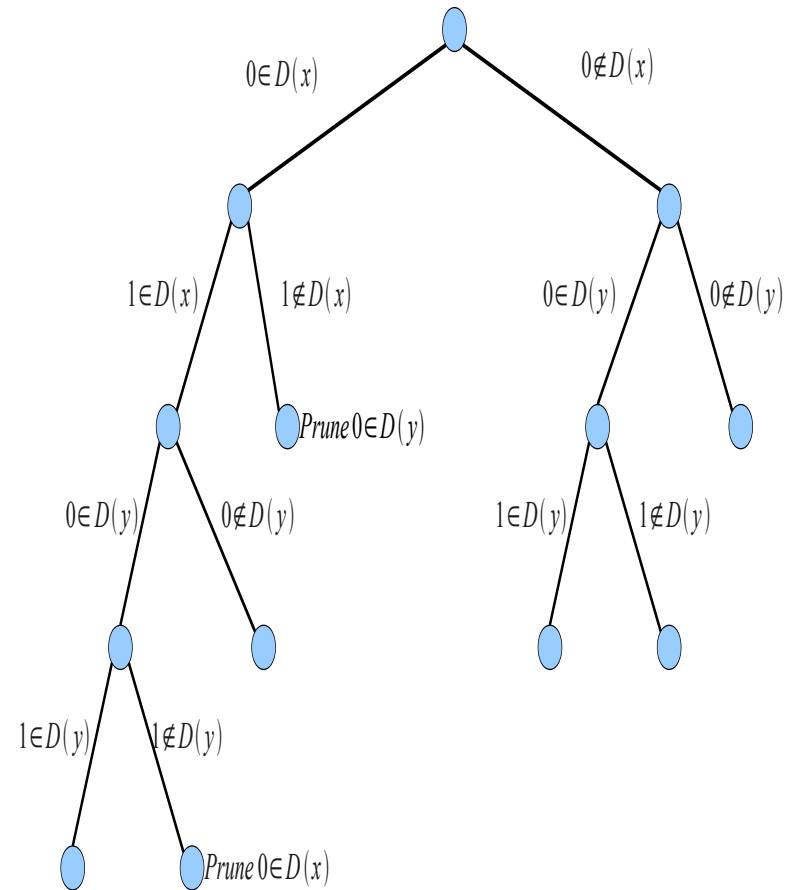
Tree for $x \vee y$, $D(x) = D(y) = \{0,1\}$

Binary decision tree

- There are only nd domain values
 - So max length of branch is nd , as is max run time
- GAC now to be run at each node in this tree
 - Instead of for every possible state of domains
 - If there are T nodes in tree then cost is
 - $O(T n d^n)$
- T can be as low as 1 for any n, d
 - E.g. domain wipeout at root

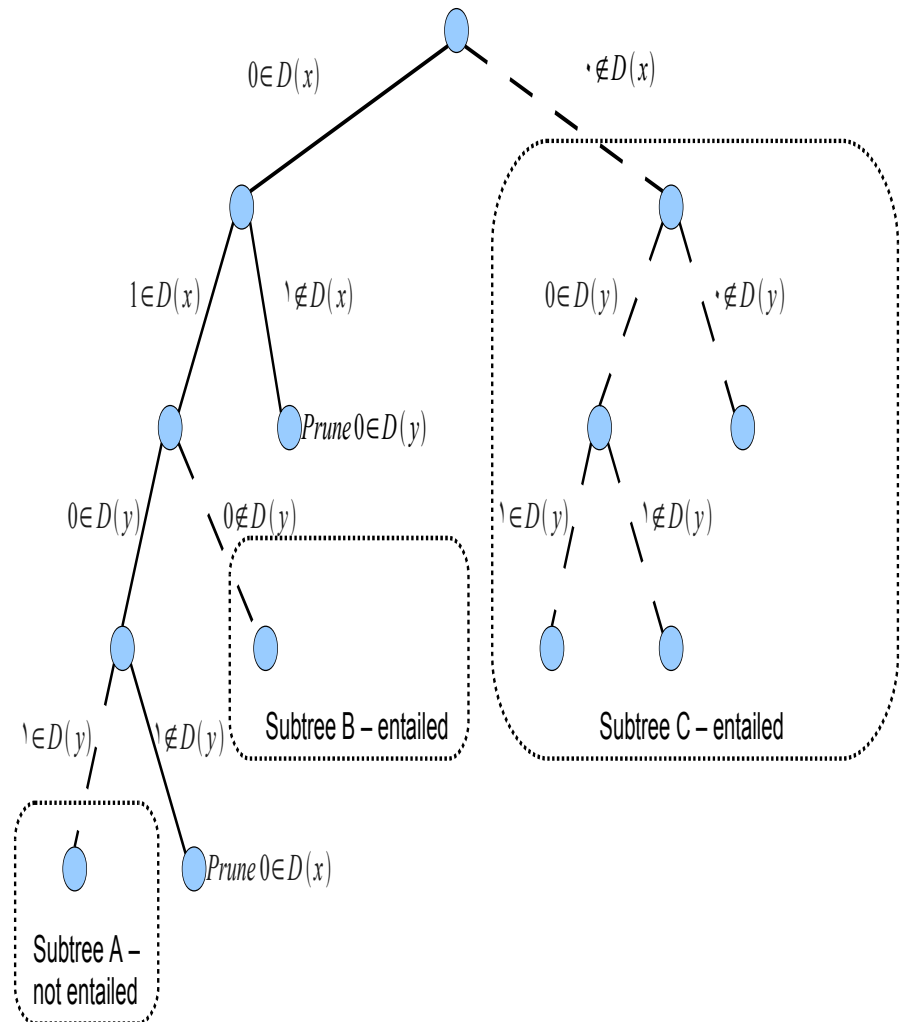
Third Idea

- Some nodes are redundant
 - Nodes where constraint is entailed
 - Other subtrees where no propagation happens
- These nodes can be turned into leaf nodes
 - Reduces space usage
 - Algorithm in paper & implemented
 - Not proved but “obviously” ok



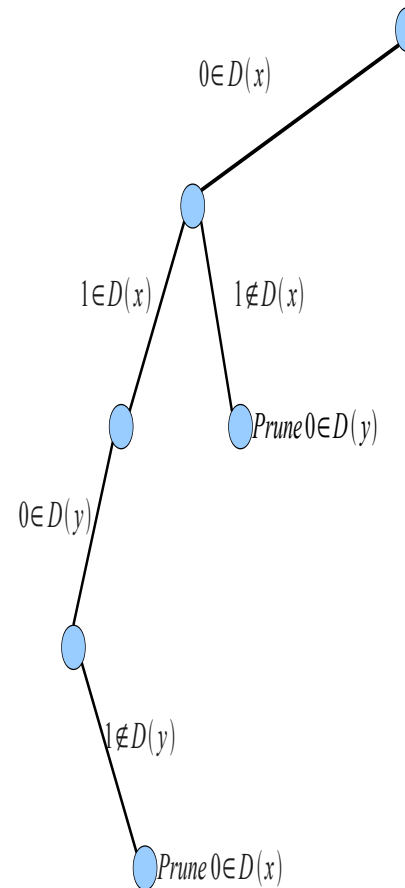
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Heuristic

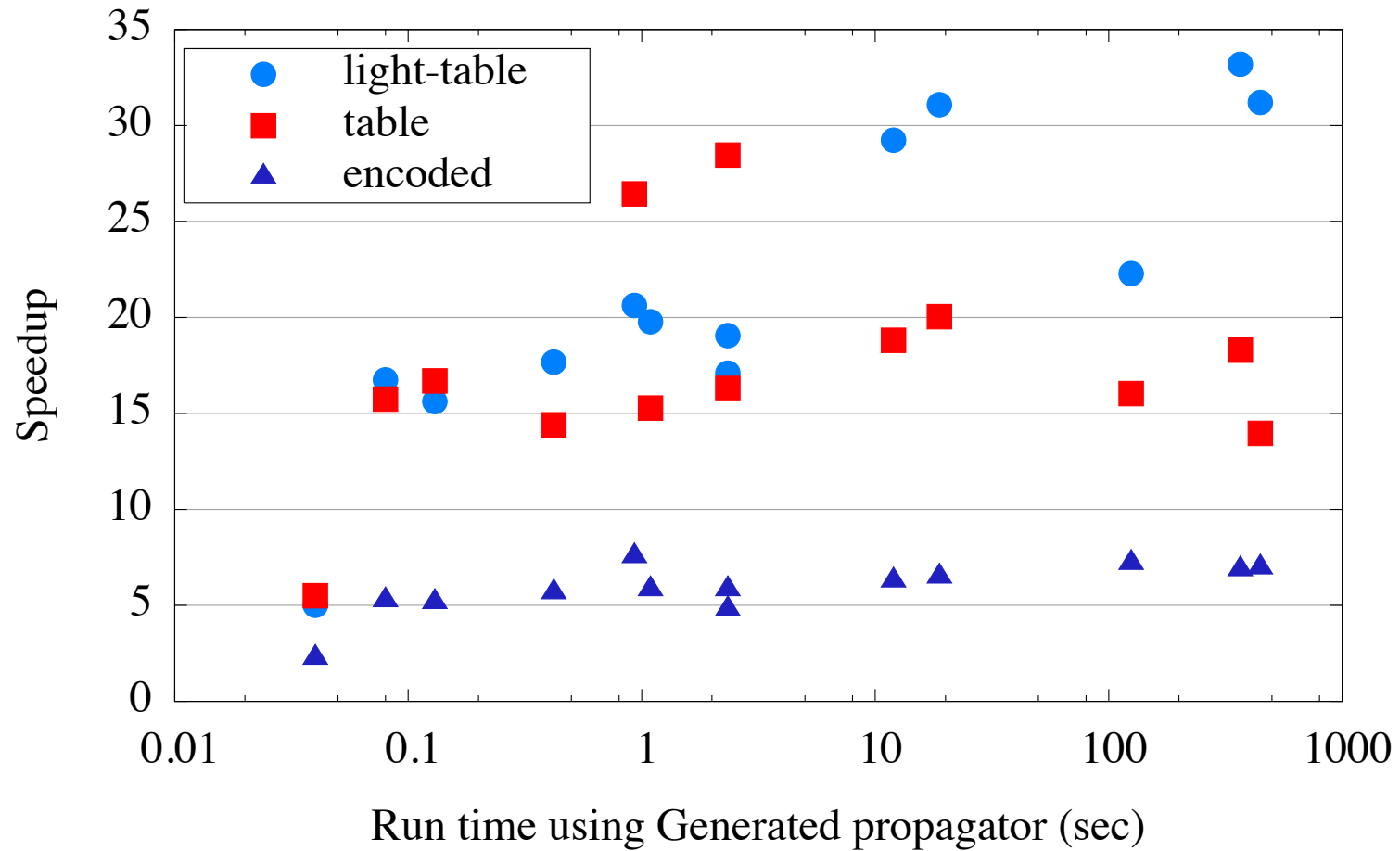
- Variable & value ordering
- Pick variable/value pair which
 - Is in maximum number of disallowed tuples
 - So that result is most likely to be entailed

Experiments

- Three case studies (great detail in paper)
 - Peg Solitaire
 - Low Autocorrelation Binary Sequences
 - Maximum Density Oscillating Still Life
- In each case constraint can be reused
 - So overheads can be amortised over all instances
- Compare against various other techniques
 - Minion GAC-table (using tries)
 - Minion “light table”
 - Encodings using other constraints

Oscillating Life Results

(overhead = 262s)

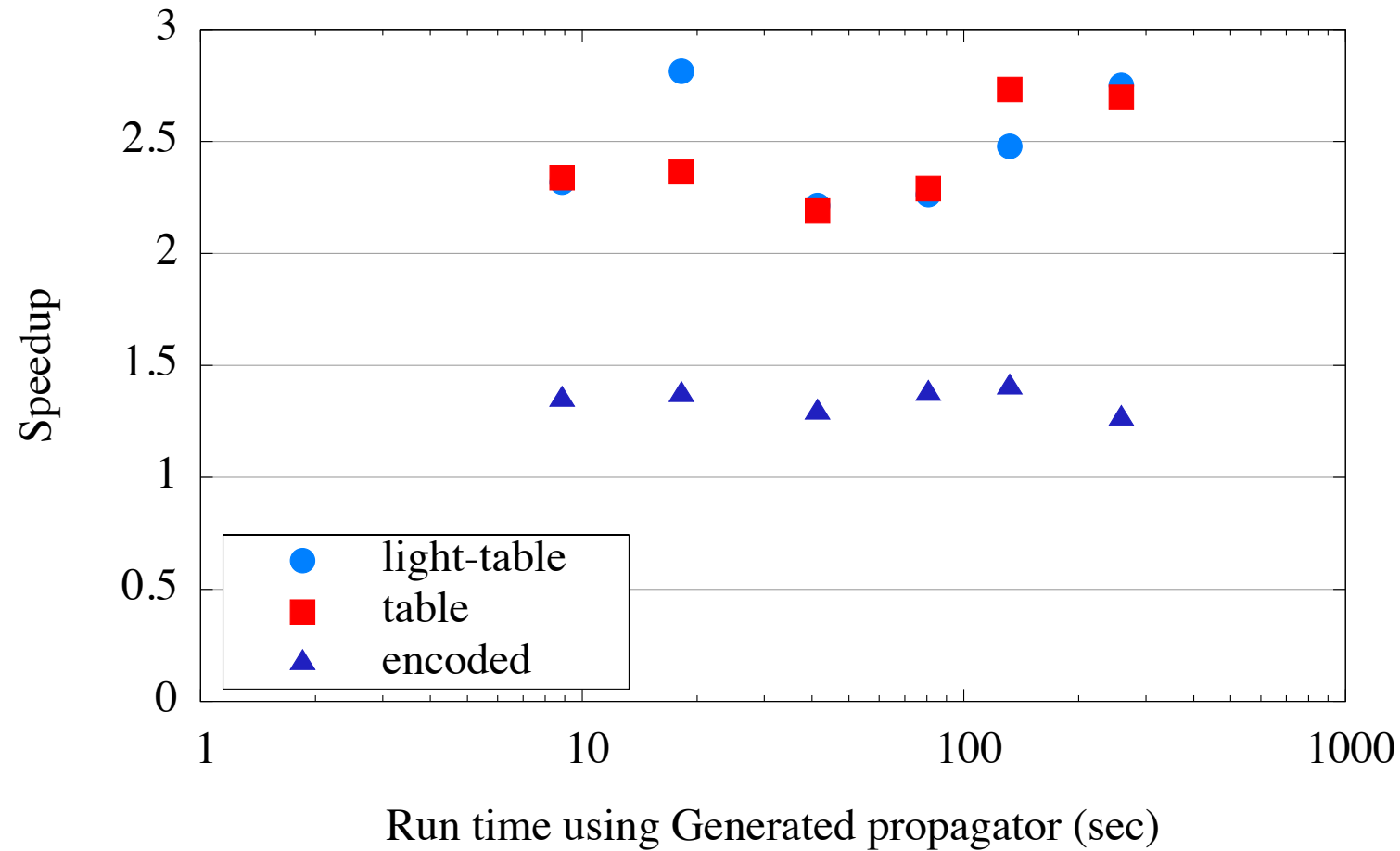


Why is this the way forward?

- **59,049**
 - states to precompute
- **564,092,290**
 - nodes in just one search

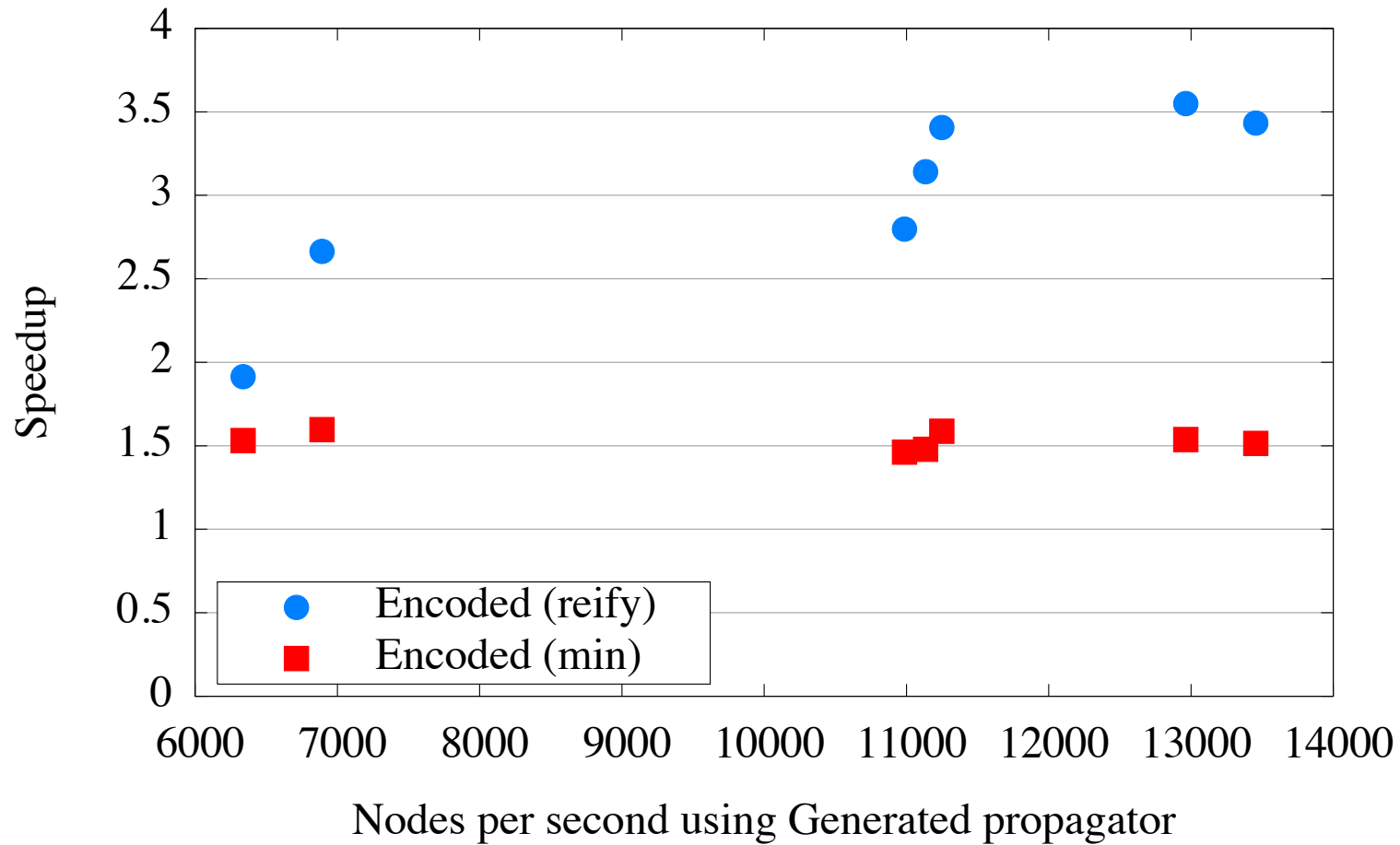
LABS Results

(overhead = 15.7s)



Peg Solitaire Results

(overhead = 16.6s)



Experiments: Key Points

- Generated *always* the fastest method
- Generated *much* faster than table/light table
 - Searching identical search space
- Generated repays overhead on hard instances
 - *And* can be reused across all instances in family
- Still faster than encodings
 - Using hand-optimised propagators!
 - And encodings need thought

Relationship to other work

- Two main bodies of related work
 - Improving GAC for table constraint
 - Lhomme/Regin 05
 - Lecoutre/Szymanek 06,
 - Cheng/Yap 06 and 10,
 - Gent/Jefferson/Miguel/Nightingale 07,
 - Katsirelos/Walsh 07
 - Lecoutre/Hemery 07
 - Constructing sets of rules in CHR to establish GAC
 - Apt/Monfroy 01
- But no work (we believe) which generates GAC propagators to be polynomial during search

Conclusions

- We aim to push GAC propagators in new direction
 - Exponential preprocessing / poly during search
 - Dominion is perfect context as we have the full details of the problem before compilation
- Shown can obtain excellent results
- Many ways work can be extended

Why is this the way forward?

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If you have been ...

... thanks for listening

Question

- Why not store the tree in a data structure instead of in code?
 - Avoids need for compiler at run time
- Yes, but compiled code runs even faster
- Even a 20% speedup can make the difference between winning and losing
- But something to consider certainly

Question

- What is the difference between entailed constraint and “no pruning”?
- Because of path to decision in binary tree
 - We *know* that some values are in domain
- E.g. $x \neq y$, original domains $D(x) = D(y) = \{0,1,2,3,4\}$
 - we might have asked
 - 0 in $D(x)$? Yes
 - 1 in $D(x)$? Yes
 - 0 in $D(y)$? Yes
 - 1 in $D(y)$? Yes
- It is now guaranteed that no propagation can occur
- But it is still possible that later we will set $x = 0$, and then propagate $y \neq 0$

Question

- Can you compute lazily?
- Yes ...
 - Obvious to think of doing GAC when propagations not cached
 - I.e. memoize propagation
 - Gets rid of exponential preprocessing
- No ...
 - Would need very different infrastructure to ours
 - Might increase cost of optimising data structure
- Obvious area of future work

Question

- Can you use BDDs/MDDs/Other data structure?
- No and Yes
- No ...
 - not just by storing the constraint
 - We still need to do all propagations in advance
- Yes ...
 - could compress data structure
 - i.e. which stores all the propagations

Question

- Is your run time really $O(nd)$?
- Is GAC-table really exponential during search?
- (Because data structure is exponential size)

- No...
 - True, input is much bigger so theoretical runtime is not what we said
- Yes...
 - After setup, exponential state is static and sits in RAM
 - In practice, access to data structure is $O(1)$

