

CROWDSOURCING INSIGHTS INTO PROBLEM STRUCTURE FOR SCIENTIFIC DISCOVERY



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Motivation: Automated Scientific Discovery





Can we use modern automated reasoning tools for scientific discovery?

Science and scientific discovery are arguably among the most impressive demonstrations of human intelligence.

We will consider two areas:

- 1) Finite Mathematics
- 2) Materials Science



I Discovery in Finite Mathematics --- A Brief Historyics



In the early days of computing, there was significant optimism with the discovery of first-order theorem proving procedures.

But, the raw size of search spaces for interesting math proved to be a daunting challenge. In a sense, the representation language (first-order logic) is too general and unconstrained to effectively explore.

Still, some success stories of automated mathematical discoveries, involving "finite mathematics." (Objects of interest are finite.)

We only consider cases of true discovery of previously unknown results (i.e. no "re-discovery").



Discoveries in Discrete Math





A) 1976 --- Four Color Thm. Four colors suffices to color a planar map. (Haken and Appel) Problem was open since 1852. Essentially solved using an exhaustive "proof by cases." (two thousand special sub-maps were analyzed. (90% human / 10% machine)

Proof is now accepted and has been replicated. Considered somewhat unsatisfying / "inelegant" from a math perspective, because the proof does not provide further insights into why the thm is true. Also, how do we know the computer proof is correct? (Do we?)

B) 1996 --- Robbins algebras are Boolean algebras. Concerns sets of axioms. (McCune) Open since 1933. Proof strategy: clever syntactic re-write of axioms. Several months of cpu time to find the re-write steps. "Semi-human" readable. (30% human / 70% machine)



C) 2014 --- Erdos Discrepancy Problem. Property of sequences of +1/-1. E.g. +1 +1 -1 +1 -1 +1 -1 ...

Does the sum (absolute value) over subsequences of this sequence stay within a certain bound, C?

Erdos conjectured (1932): If sequence gets long enough "discrepancy" exceeds any value of C. Lisitsa and Konev (2014) showed every sequence of 1161 or longer exceeds bound of 2 (i.e. C = 2). But, remarkably, they also found a +1/-1 sequence of 1160 elements that stayed within bound of 2. (10% human / 90% machine)

Obtained with Boolean **Satisfiability** solver (20 hrs). Proof trace of **13 gigabytes**. Longest proof in history. Proof **verified** by proof checker. (Contrast two previous results.)





Each result represents a clear new mathematical advance. But, unfortunately, approaches do not provide much insight into the underlying problem domain.

For example, can we understand how certain complex objects (e.g. 1160 long +1/-1 sequence or coloring of sub-map) can be constructed directly using a human-understandable strategy?



Example Domain:



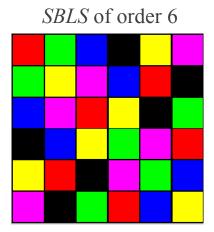


The Spatially-balanced Latin square (SBLS) problem

Problem Definition:

An SBLS of order n is an $n \times n$ square grid in which:

Each symbol appears exactly once in each row and column (*Latin square* structure; multiplication table of quasi-group).





Our Challenge Domain:

ICS

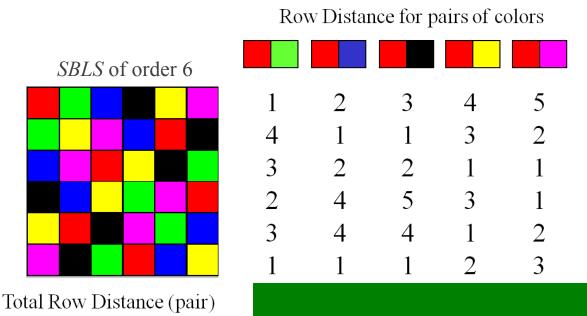


The Spatially-Balanced Latin Square (SBLS) problem

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An SBLS of order n is an $n \times n$ square grid in which:

- Each symbol appears exactly once in each row and column (*Latin square* structure; multiplication table of quasi-group).
- The average distance (column-wise) of a pair of symbols is the same for any pair (Balanced structure).





Average Row Distance (pair)

The Spatially-Balanced Latin Square (SBLS) problem

SBLS of order 6

Row Distance for pairs of colors

	1	2	3	4	5	
	4	1	1	3	2	
	3	2	2	1	1	
	2	4	5	3	1	
	3	4	1	1	2	
	1	1	2	2	3	
Total Row Distance (pair)	14	14	14	14	14	<u> </u>
Average Row Distance (pair)	2.33	2.33	2.33	2.33	2.33	

14 (avg. 2.33) for all pairs. Use in "experimental design."

In Computational Sustainability: to limit use of fertilizers.

Designs used by farmers in New York State!

Finding SBLS is very hard.





Approach	Order	Time (s)	Reference
Constraint Programming (CP)	9	241	[Gomes and Sellmann, CP'04]
IDWalk (metaheuristic)	9	4.5	[Neveu et al., CP'04]
Self-symmetry-based Streamlined CP	14	5,434	[Gomes and Sellmann, CP'04]
Composition-based Streamlined CP	18	107K	[Gomes and Sellmann, CP'04]
Streamlined Local Search	35	1.2M	[Smith et al., IJCAI'05]

Approach: use lots of problem specific structure to guide search.

Earlier known designs, only up to order 6.

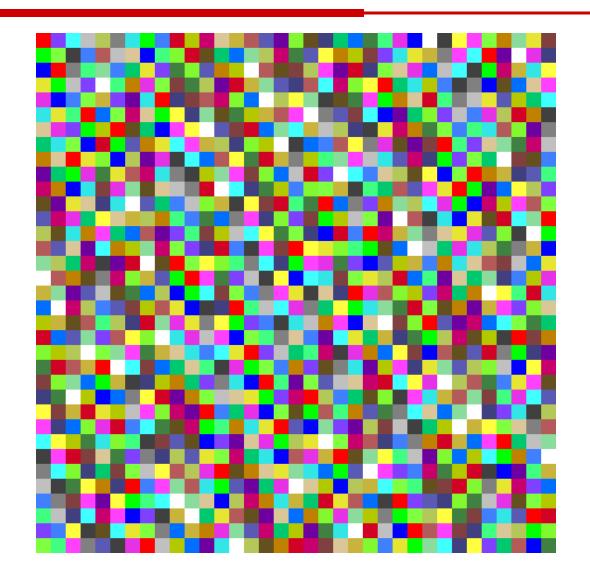
The balancing condition (of each pair!) makes the problem *very* hard.



Largest ever found 35x35 --- 330 hrs cpu time.





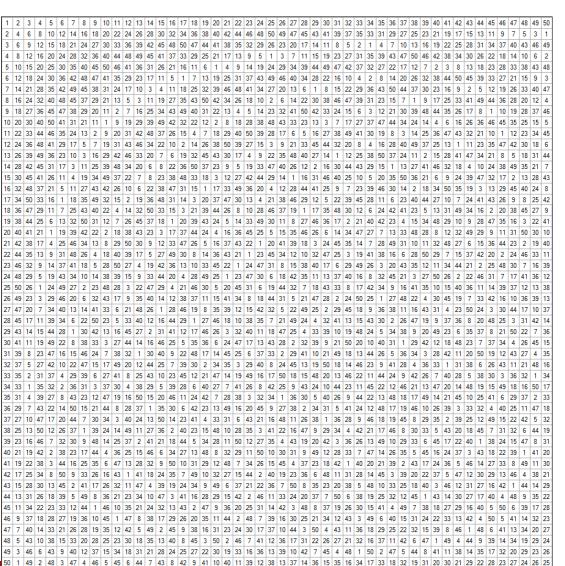




This work: 50x50 in << 1 ms







How?

(Nothing special about 50x50.

We can do

10Kx10K

in a few seconds.)

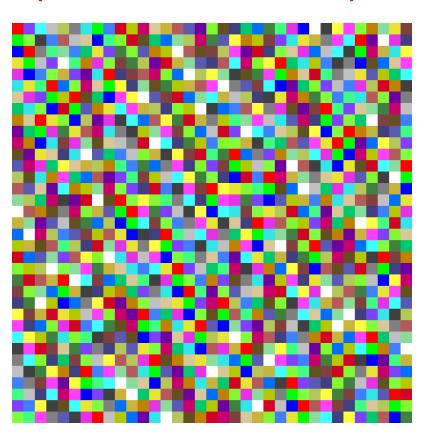


Gedanken Experiment:

(here)

Given that a constructive algorithm (given any size n) exists, how would you find it? Hwk...

(Can use SAT solver.)



Outline





- Motivation
- Example Domain
- Proposed Framework
 - Overview of Streamlined Search
 - Taking advantage of Human Insights
 - Formal Description and Overview
 - Human-guided Streamlined Search
- Application to the Spatially-balanced Latin square problem
- Application to the Weak Schur Number problem
- Conclusions and Future work



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Proposed Framework:

Overview of Streamlined Combinatorial Search





Goal:

Exploit the structure of some solutions to boost the effectiveness of the propagation mechanisms.

Underlying Observation:

When one insists on maintaining the **full solution set**, there is a **hard practical limit** on the effectiveness of **constraint propagation** methods. Often, there is **no compact representation** for all the solutions.

Underlying Conjecture:

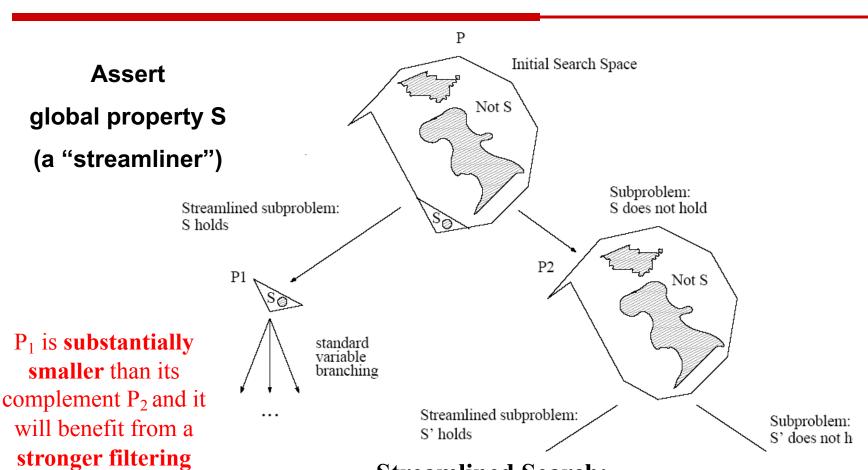
For many intricate **combinatorial problems** – if **solutions exist** – there will often be (highly) **regular ones**. **Can we find such solutions?**



Overview of Streamlined Search







Streamlined Search:

Strong **branching mechanisms** (by adding constraints based on **structure properties**) at **high levels** of the search tree.



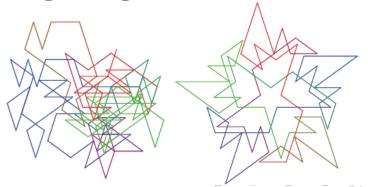
thereafter.

Taking advantage of Human Insights



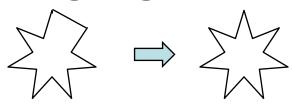


Recognizing Patterns and Regularities:



[Source: Marijn J.H. Heule, 2009, in work on van der Waerden numbers]

Correcting Irregularities:



Generalizing / Formalizing Regularities:

1	2	3
3	1	2
2	3	1



1 2 3 4 4 1 2 3 3 4 1 2 2 3 4 1

Cyclic Latin square of order 3

Cyclic Latin square of order 4



Proposed Framework:

Formal Description and Overview





```
\mathcal{O} \leftarrow \emptyset:
                                                     // Conjectured streamliners
\Gamma \leftarrow \emptyset:
                                                             // Search streamliners
                                                                 // Search parameter
\rho \leftarrow \rho_0;
S \leftarrow \emptyset:
                                                                   // Solutions found
\tau \leftarrow false;
                                                                        // Timeout flag
repeat
      Solve(P_{\rho}, \Gamma, t) \to (S', \tau); // Search for new solutions
      if \mathcal{S}' \cap \mathcal{S} \neq \emptyset then
            \mathcal{S} \leftarrow \mathcal{S} \cup \mathcal{S}':
                                     // Case 1: successful search
            Analyze(S) \rightarrow \mathcal{O}'; // Conjecture new streamliners
            \mathcal{O} \leftarrow \mathcal{O} \cup \mathcal{O}':
            \rho \leftarrow \rho + 1;
      else if \tau is true then
            Select \Gamma' \subset \mathcal{O}; // Case 2: timed-out failed search
            \Gamma \leftarrow \Gamma \cup \Gamma':
                                                      // Strengthen streamliners
      else
            Select \Gamma' \subseteq \Gamma; // Case 3: exhaustive failed search
            \Gamma \leftarrow \Gamma \setminus \Gamma';
                                                           // Weaken streamliners
            \rho = \max\{\rho : \mathcal{S}(\Gamma) \cap \mathcal{S}(P_{\rho}) \neq \emptyset\} + 1;
            Select \Gamma'' \subset \Gamma'; // Find next parameter of interest
            \mathcal{O} \leftarrow \mathcal{O} \setminus \Gamma''; // Drop unpromising streamliners
until \mathcal{O} = \emptyset:
```

Algorithm: Discover-Construction procedure for a given problem P, with parameter set ρ and timeout t.



Overview Approach --- Mimics Alg Discovery





```
\mathcal{O} \leftarrow \emptyset:
                                                       // Conjectured streamliners
\Gamma \leftarrow \emptyset:
                                                               // Search streamliners
                                                                   // Search parameter
\rho \leftarrow \rho_0;
\mathcal{S} \leftarrow \emptyset:
                                                                     // Solutions found
\tau \leftarrow false;
                                                                           // Timeout flag
repeat
      Solve(P_{\rho}, \Gamma, t) \to (S', \tau); // Search for new solutions
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             \mathcal{O} \leftarrow \mathcal{O} \setminus \Gamma'';
                                            // Drop unpromising streamliners
until \mathcal{O} = \emptyset:
```

Algorithm: Discover-Construction procedure for a given problem P, with parameter set ρ and timeout t.

- 1 Analyze smaller size solutions, and conjecture potential regularities in the solutions. (Human insight.)
- 2 Validate through **streamlining** the observed regularities.
- 3 If the streamlined search does not give a larger size solution, the proposed regularity is quite likely accidental and one looks for a new pattern in the small scale solutions.
- 4 Otherwise, one proceeds by generating a number of **new solutions** that all contain the proposed **structural regularity** and are used to expand the solution set and to **reveal new regularities**.







Overview on the SBLS problem

Search Parameters

$$n=3$$

$$\Gamma={}$$

Conjectured Streamliners

Start with first order of interest (n=3) and no streamliners $(\Gamma=\{\})$









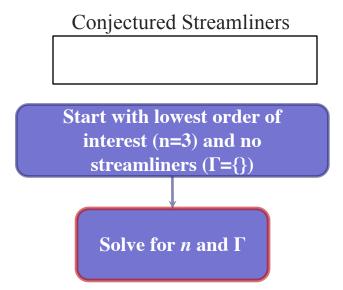


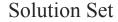
Overview on the SBLS problem

Search Parameters

n=3

 $\Gamma = \{\}$



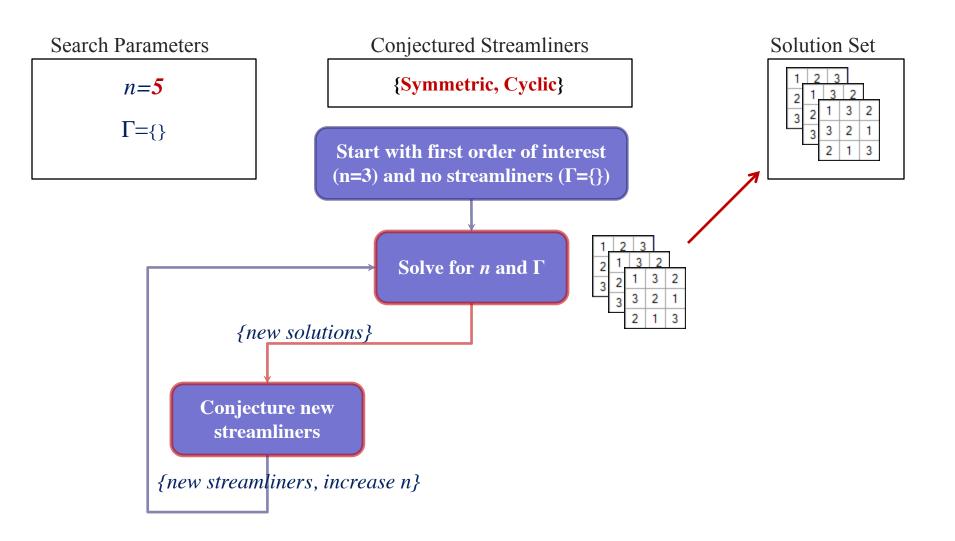










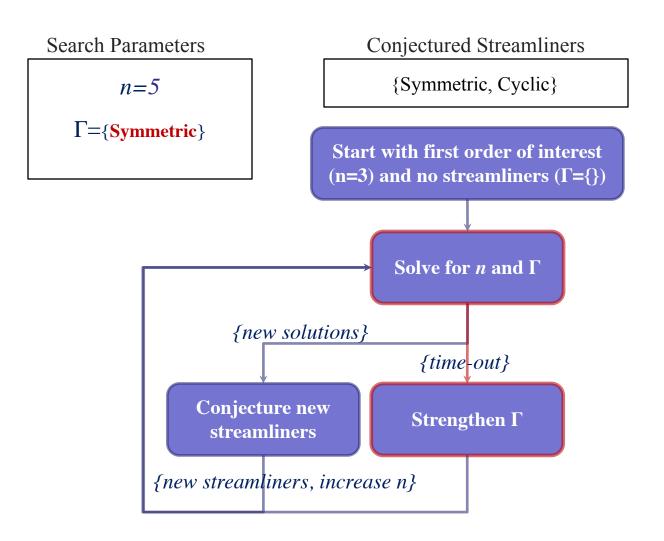




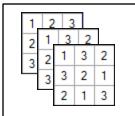




Overview on the SBLS problem

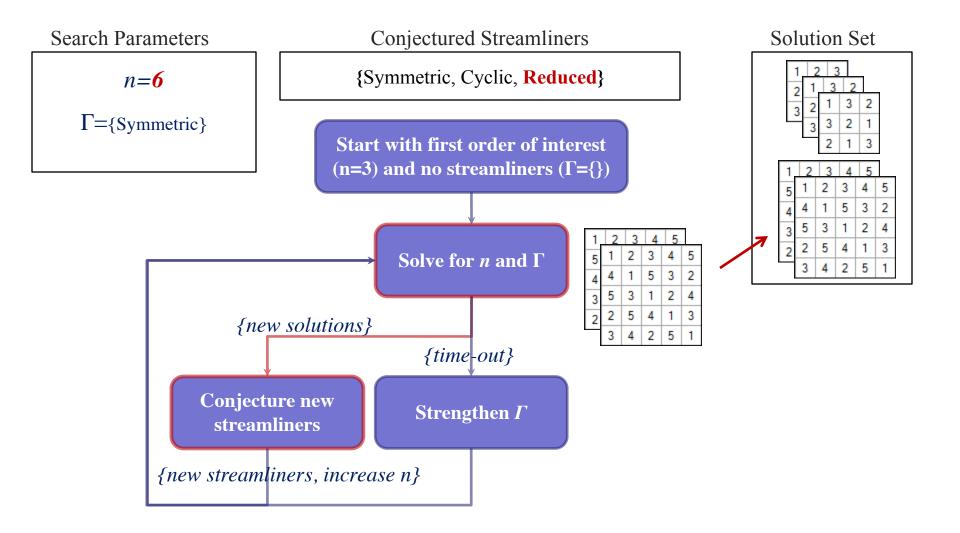


Solution Set





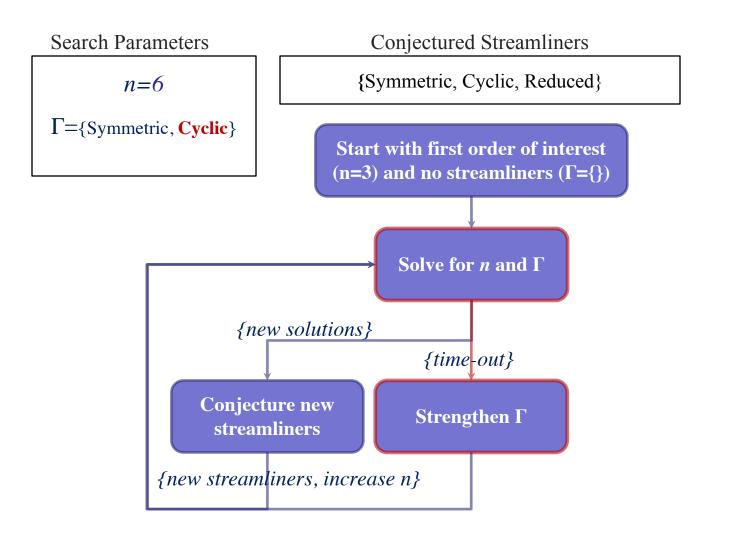


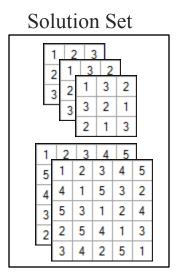






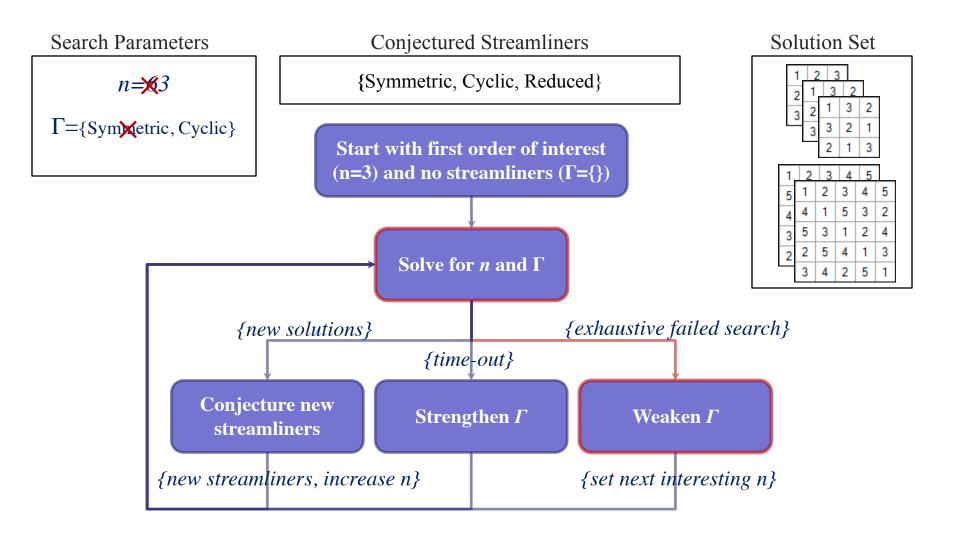








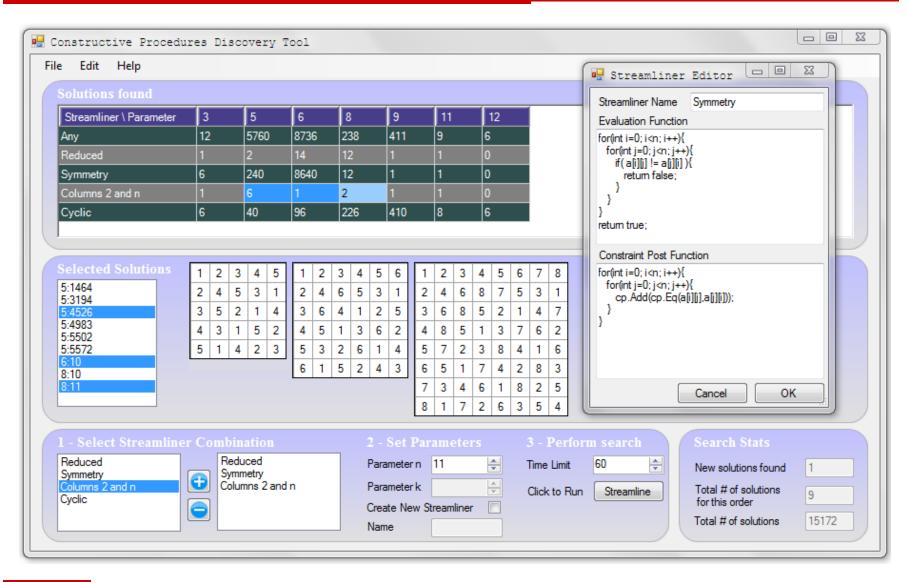








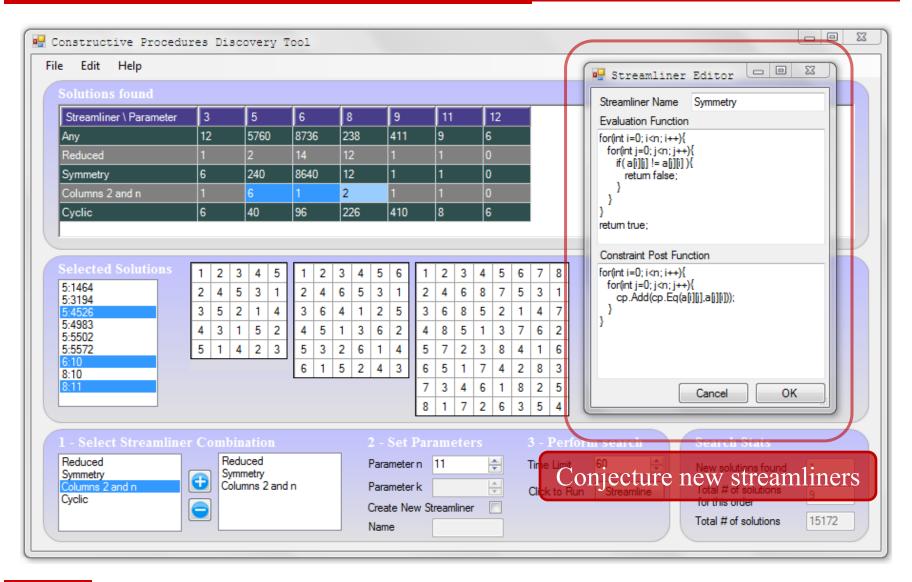








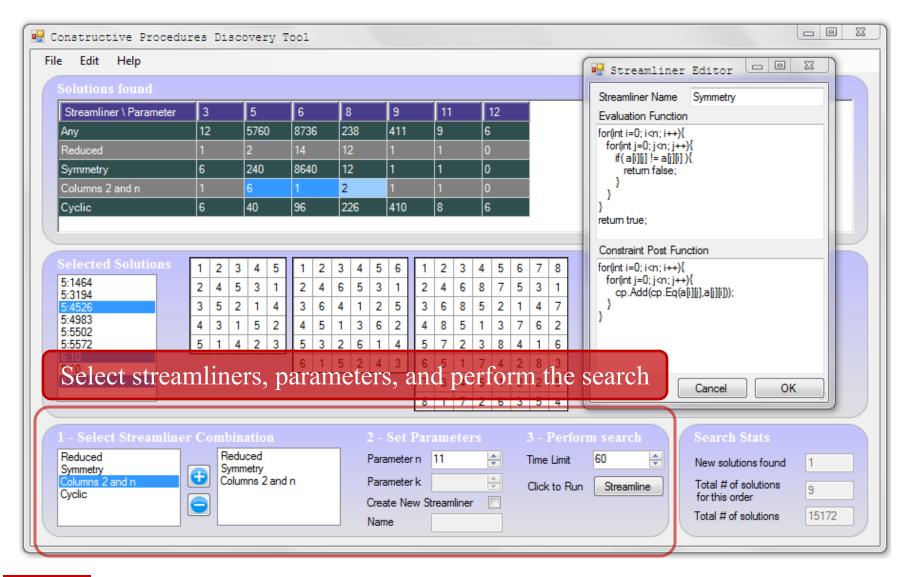








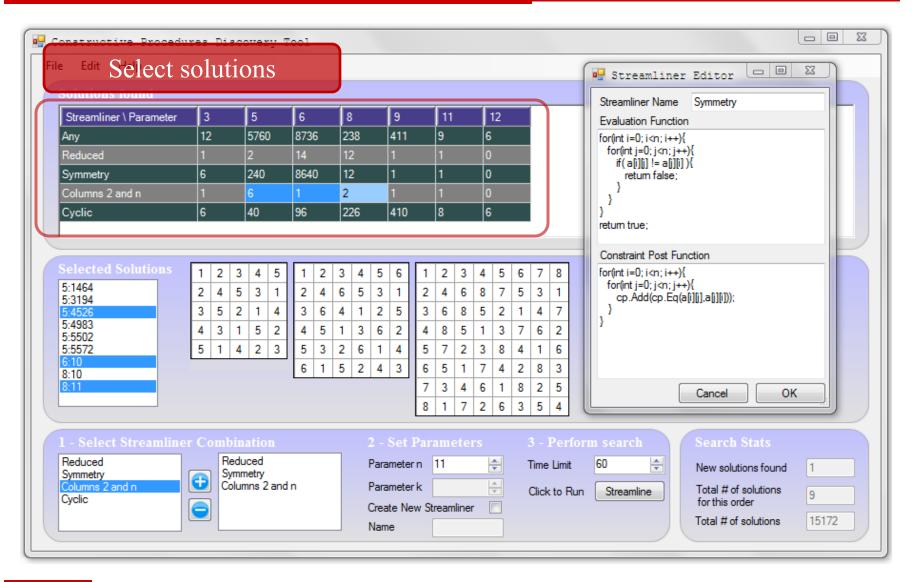








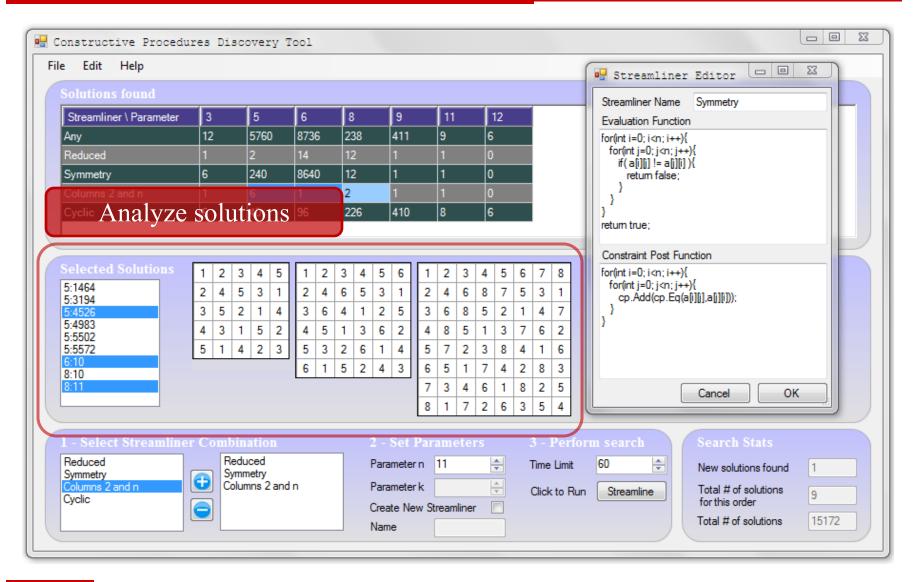














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 - Successful Streamliners
 - Constructive Procedure 1
 - Constructive Procedure 2
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Successful Key Streamliners:

{Diagonal symmetry, Reduced form, Assignments of columns 2 and *n*, Multiples of *i* in row *i*, Second sequence decreasing}

Streamliners	5	6	8	9	11	14
$\Gamma_1 = \emptyset$	5760	15878	-	-	-	-
$\Gamma_2 = \Gamma_1 \cup \{Symmetric\}$	240	8447	714	43	-	-
$\Gamma_3 = \Gamma_2 \cup \{Reduced\}$	2	14	14	51	-	-
$\Gamma_4 = \Gamma_3 \cup \{Columns\ 2 \& n\}$	1	1	2	1	1	-
$\Gamma_5 = \Gamma_4 \cup \{Multiples \ of \ i\}$	1	1	2	1	1	1

Fig: Number of SBLSs generated in 60 seconds, by order and streamliners (Bold indicates exhaustive search).



Application to the SBLS problem: Construction 1





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

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

1	2	3	4	5	6	7	8	9	10	11
2	4	6	8	10	11	9	7	5	3	1
3	6	9	11	8	5	2	1	4	7	10
4	8	11	7	3	1	5	9	10	6	2
5	10	8	3	2	7	11	6	1	4	9
6	11	5	1	7	10	4	2	8	9	3
7	9	2	5	11	4	3	10	6	1	8
8	7	1	9	6	2	10	5	3	11	4
9	5	4	10	1	8	6	3	11	2	7
10	3	7	6	4	9	1	11	2	8	5
11	1	10	2	9	3	8	4	7	5	6

1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	4	6	8	10	12	14	13	11	9	7	5	3	1
3	6	9	12	14	11	8	5	2	1	4	7	10	13
4	8	12	13	9	5	1	3	7	11	14	10	6	2
5	10	14	9	4	1	6	11	13	8	3	2	7	12
6	12	11	5	1	7	13	10	4	2	8	14	9	3
7	14	8	1	6	13	9	2	5	12	10	3	4	11
8	13	5	3	11	10	2	6	14	7	1	9	12	4
9	11	2	7	13	4	5	14	6	3	12	8	1	10
10	9	1	11	8	2	12	7	3	13	6	4	14	5
11	7	4	14	3	8	10	1	12	6	5	13	2	9
12	5	7	10	2	14	3	9	8	4	13	1	11	6
12 13	3	10	6	7	9	4	12	1	14	2	11	5	8
14	1	13	2	12	3	11	4	10	5	9	6	8	7



Application to the SBLS problem: Construction 1





1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	4	6	8	10	12	14	13	11	9	7	5	3	1
3	6	9	12	14	11	8	5	2	1	4	7	10	13
4	8	12	13	9	5	1	3	7	11	14	10	6	2
5	10	14	9	4	1	6	11	13	8	3	2	7	12
6	12	11	5	1	7	13	10	4	2	8	14	9	3
7	14	8	1	6	13	9	2	5	12	10	3	4	11
8	13	5	3	11	10	2	6	14	7	1	9	12	4
9	11	2	7	13	4	5	14	6	3	12	8	1	10
10	9	1	11	8	2	12	7	3	13	6	4	14	5
11	7	4	14	3	8	10	1	12	6	5	13	2	9
12	5	7	10	2	14	3	9	8	4	13	1	11	6
13	3	10	6	7	9	4	12	1	14	2	11	5	8
14	1	13	2	12	3	11	4	10	5	9	6	8	7



Approach reveals patterns. Final step is figuring out boundary conditions.



Application to the *SBLS* problem: Construction 1





1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	4	6	8	10	12	14	13	11	9	7	5	3	1
[3]	6	9	12	14	11	8	5	2	[1	4	7	10	13
4	8	12	13	9	5	1	3	7	11	14	10	6	2
5	10	14	9	4	1	6	11	13	8	3	2	7	12
6	12	11	5	1	7	13	10	4	2	8	14	9	3
7	14	8	1	6	13	9	2	5	12	10	3	4	11
8	13	5	3	11	10	2	6	14	7	1	9	12	4
9	11	2	7	13	4	5	14	6	3	12	8	1	10
10	9	1	11	8	2	12	7	3	13	6	4	14	5
11	7	4	14	3	8	10	1	12	6	5	13	2	9
12	5	7	10	2	14	3	9	8	4	13	1	11	6
13	3	10	6	7	9	4	12	1	14	2	11	5	8
14	1	13	2	12	3	11	4	10	5	9	6	8	7

```
for row i = 1, \ldots, N do
    k = 1;
                                         // Sequence number
   j = 1;
                                            // Column index
    a_{i,j} = i;
                                     // First symbol of row i
    while j < N do
        if k is odd then
                                            // Odd sequence
            while a_{i,j} + i \leq N and j < N do
                a_{i,j+1} = a_{i,j} + i;
                j = j + 1;
        else
                                            // Even sequence
            while a_{i,j} - i \ge 1 and j < N do
                a_{i,j+1} = a_{i,j} - i;
                j = j + 1;
        if j < N then
                                          // Switch sequence
            if k is odd then
                a_{i,i+1} = 2N + 1 - i - a_{i,i};
            else
               a_{i,j+1} = i - a_{i,j};
            k = k + 1:
            j = j + 1;
```

a+i a a-i

Algorithm: SBLS-sequence procedure for SBLS of order N, when 2N+1 is prime.



Application to the *SBLS* problem: Construction 1





1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	4	6	8	10	12	14	13	11	9	7	5	3	1
3	6	9	12	14	11	8	5	2	1	4	7	10	13
4	8	12	13	9	5	1	3	7	11	14	10	6	2
5	10	14	9	4	1	6	11	13	8	3	2	7	12
6	12	11	5	1	7	13	10	4	2	8	14	9	3
7	14	8	1	6	13	9	2	5	12	10	3	4	11
8	13	5	3	11	10	2	6	14	7	1	9	12	4
9	11	2	7	13	4	5	14	6	3	12	8	1	10
10	9	1	11	8	2	12	7	3	13	6	4	14	5
11	7	4	14	3	8	10	1	12	6	5	13	2	9
12	5	7	10	2	14	3	9	8	4	13	1	11	6
13	3	10	6	7	9	4	12	1	14	2	11	5	8
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```
for row i = 1, \ldots, N do
    k = 1;
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    while j < N do
        if k is odd then
                                            // Odd sequence
            while a_{i,j} + i \leq N and j < N do
                a_{i,j+1} = a_{i,j} + i;
                j = j + 1;
        else
                                            // Even sequence
            while a_{i,j} - i \ge 1 and j < N do
                a_{i,j+1} = a_{i,j} - i;
                j = j + 1;
        if j < N then
                                          // Switch sequence
            if k is odd then
                a_{i,i+1} = 2N + 1 - i - a_{i,i};
            else
               a_{i,j+1} = i - a_{i,j};
            k = k + 1;
            j = j + 1;
```

a a+i a a-i

Algorithm: SBLS-sequence procedure for SBLS of order N, when 2N+1 is prime.

Proof of Correctness in [R. Le Bras, et al., *Polynomial Time Construction for Spatially Balanced Latin Squares*, 2012]



Application to the *SBLS* problem: Construction 2





1	2	6	3	9	7	13	4	11	10	12	8	5	14
2	3	7	4	10	8	14	5	12	11	13	9	6	1
3	4	8	5	11	9	1	6	13	12	14	10	7	2
4	5	9	6	12	10	2	7	14	13	1	11	8	3
5	6	10	7	13	11	3	8	1	14	2	12	9	4
6	7	11	8	14	12	4	9	2	1	3	13	10	5
7	8	12	9	1	13	5	10	3	2	4	14	11	6
8	9	13	10	2	14	6	11	4	3	5	1	12	7
9	10	14	11	3	1	7	12	5	4	6	2	13	8
10	11	1	12	4	2	8	13	6	5	7	3	14	9
11	12	2	13	5	3	9	14	7	6	8	4	1	10
12	13	3	14	6	4	10	1	8	7	9	5	2	11
13	14	4	1	7	5	11	2	9	8	10	6	3	12
14	1	5	2	8	6	12	3	10	9	11	7	4	13

```
c_{1,1}=1;
                         // Generate 1st row of the conjugate
for column j = 2, ..., N do // Observed pattern 1, 2, 4, ...
    if 2c_{1,j-1} \leq N then
       c_{1,j} = 2c_{1,j-1};
    else
        c_{1,i} = 2N + 1 - 2c_{1,i-1};
for row i = 2, \dots, N do
                                          // Subsequent rows
    c_{i,1} = c_{i-1,N};
                               // Shifted version of previous
    for column j = 2, \dots, N do
        c_{i,j} = c_{i-1,j-1};
for row i = 1, ..., N do // Generate SBLS from conjugate
    for column j = 1, \dots, N do
        a_{i,c_{i,j}}=j;
```

Algorithm: SBLS-Cyclic procedure.



Observation I





Incremental approach in terms of size is necessary:

Need to reveal structure by generating increasingly larger solutions using increased streamlining (and thus, structure).

True solution structure only becomes visible around order 14.

At that size

- (1) Can't generate any solution of that size without strong streamlining,
- (2) But, even if we could, arbitrary solutions will not show structure.



Observation II





We conjecture that many hard combinatorial design problems have effective constructive procedures. (E.g., quasigroup existence problems, code design, Ramsey graphs (numbers), van der Waerden, Schur numbers etc.).

So far, such problems have been tackled with combinatorial search (SAT solvers, equational thm. provers) on a per instance basis. Does resolve open questions in combinatorial design (e.g. new lower-bounds) but does not reveal the general solution structure.

E.g. Slaney et al. 1993, Herwig et al. 2007, Heule and Walsh 2010 Eliahou et al. 2012, Fujita et al. 2013.



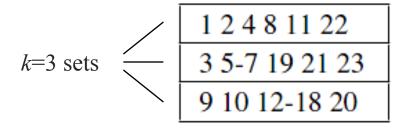
Example: the Weak Schur problem





Problem Definition:

- A set is *(weakly)* sum free if for any two *(distinct)* elements of this set, their sum does not belong to the set.
- The *Weak Schür Number* of order k, WS(k), is the largest integer n for which there exists a partition of [1,n] into k weakly sum-free sets.



Each of the 3 sets is such that, for any 2 elements of the set, their sum does not belong to the same set.

Fig: Partition of [1,23] into 3 weakly sumfree sets, proving $WS(3) \ge 23$



Application to the Weak Schur problem





Best known lower bounds:

Approach	WS(5)	WS(6)	Reference
(not disclosed)	196	-	[G.W. Walker, AMM'50]
Theoretical bound (not proved)	188	554	[J.H. Braun, AMM'50]
SAT	196	572	[Eliahou et al., Computers & Math Applications'12]
Multi-level Tabu-Search	196	574	[Fonlupt et al., EA'11]
SAT (no certificate)	196	575	[Eliahou et al., Computers & Math Applications'12] (revised)

New: $WS(6) \ge 581$







Successful Key Streamliners:

{Ordered sets, constrained minimum of each set, partial assignments, sequences of consecutive integers, sequence interleaving}



Application to the Weak Schur problem





Successful Key Streamliners:

{Ordered sets, constrained minimum of each set, partial assignments, sequences of consecutive integers, sequence interleaving}

1 2 4 8 11 22 25 50 63 68 139 149 154 177 182 192 198 393 398 408 413 436 450 455 521 526 540 563 568 578
3 5-7 19 21 23 51-53 64-66 136-138 150-152 179-181 193-195 395-397 409-411 438-440 451-453 523-525 536-538 565-567 579-581
9 10 12-18 20 54-62 140-148 183-191 399-407 441-449 527- 535 569-577
24 26-49 153 155-176 178 412 414-435 437 539 541-562 564
67 69-135 454 456-520 522
196 197 199-392 394

Partition of [1,581] into 6 weakly sumfree sets, proving $WS(6) \ge 581$.

(le Bras, Gomes,

Selman AAAI 2012)

Incrementally obtained, although **not** an example of a **fully constructive procedure** yet, any progress on Schur numbers is quite **significant** given their long history.



Summary





- Structure discovery through a general framework that integrates streamlined combinatorial search with human insight and intuition in an iterative approach to discover efficient constructive procedures.
- Provides the **first constructive procedures** for the **Spatially-Balanced Latin Square** problem.
- Also, improves the best known **lower bound** for the *Weak Schur Number* problem, and, most recently, for "Graceful graphs."





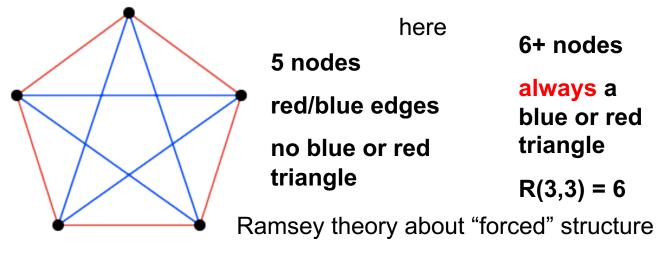


Extensions **crowd-source** the search for regularities in the solution sets (on Mechanical Turk).

And much, much more ambitiously...

- --- Ramsey numbers (Erdos), and
- --- possibly, new types of algorithms.





here

5 nodes

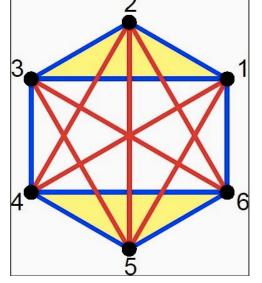
red/blue edges

no blue or red triangle

6+ nodes

always a blue or red triangle

$$R(3,3) = 6$$



"unavoidable structure in our universe"

R(3,3) = 6

R(4,4) = 18

R(5,5) = ?

Erdos: "Imagine an alien force, vastly more powerful than us landing on Earth and demanding the value of R(5, 5) or they will destroy our planet." [hmm?]

In that case, we should marshal all our computers and all our mathematicians and attempt to find the value. ...

But suppose, instead, that they asked for R(6, 6), then we should attempt to destroy the aliens".

The key will be to find the hidden structure in the solution space.

Part II: Crowdsourcing for Backdoor Variables: Materials Discovery Domain



Briefly!

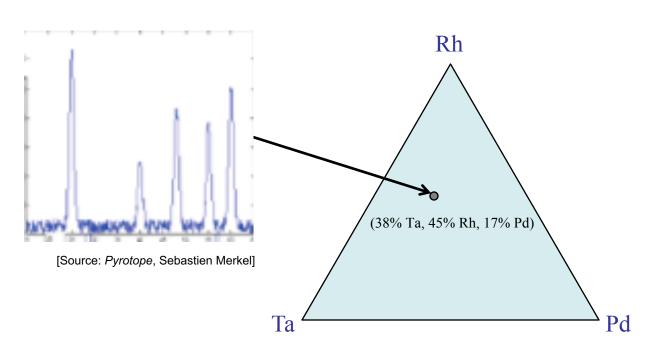


Warning: Real-world, noisy data ahead



Motivation

Study of new materails: Identifying crystalline structure using **X-Ray Diffraction patterns**



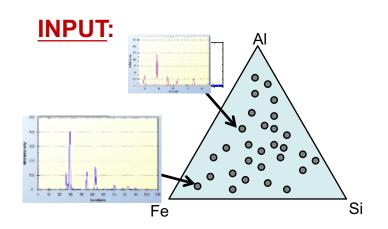


With Caltech, working on screening 1 million samples per day.

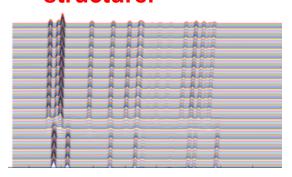
Phase Map Identification Problem





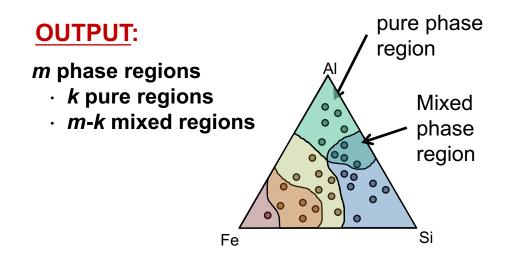


Collection of X-Ray diffraction Patterns. **Noisy and hidden** structure.



Additional Physical Requirements:

- Phase Connectivity
- Mixtures of ≤ 3 pure phases
- Peaks shift by ≤ 15% within a region
 - Continuous and Monotonic
- Small peaks might be discriminative
- Peak locations matter,
 more than peak intensities





Complex Full Constraint Optimization Model

Variables	Description
p_{ki}	Normalizing element for phase k in pattern P_i
a_{ki}	Whether phase k is present in pattern P_i
q_k	Set of normalized peak locations of phase B_k

$$(a_{ki} = 0) \iff_K (p_{ki} = 0) \qquad \forall \ 1 \le k \le K, 1 \le i \le n$$
 (1)

$$1 \le \sum_{i=1}^{K} a_{si} \le M \qquad \forall \ 1 \le i \le n \tag{2}$$

$$(p_{ki} = j) \Rightarrow (q_k \subseteq r_{ij}) \qquad \forall \ 1 \le k \le K, 1 \le i \le n, 1 \le j \le |P_i|$$
 (3)

$$(p_{ki} = j \land \sum_{s=1}^{K} a_{si} = 1) \Rightarrow (r_{ij} \subseteq q_k)$$

$$\forall 1 \le k \le K, 1 \le i \le n, 1 \le j \le |P_i| \qquad (4)$$

$$(p_{ki} = j \land p_{k'i} = j' \land \sum_{s=1}^{K} a_{si} = 2)$$

$$\Rightarrow (member(r_{ij}[j''], q_k) \lor member(r_{ij'}[j''], q_{k'}))$$

$$\forall 1 \le k, k' \le K, 1 \le i \le n, 1 \le j, j', j'' \le |P_i|$$
 (5)

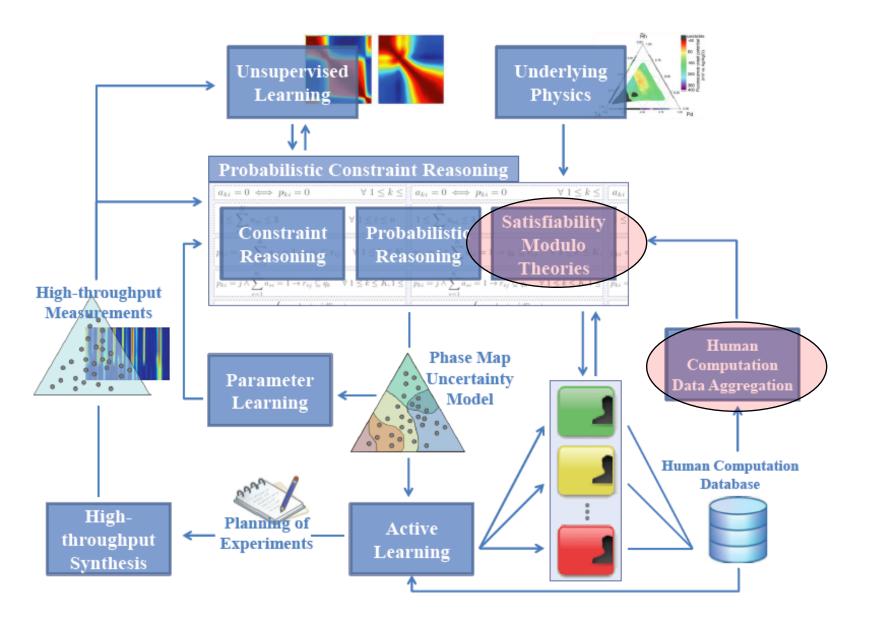
$$(p_{ki} = j) \Rightarrow (p_{ki'} \neq j') \qquad \forall \ 1 \leq k \leq K, (i, j, i', j') \in \Phi$$

$$\Phi = \{(i, j, i', j') \mid \frac{P_i[j]}{P'_i[j']} < 1/\delta \lor \frac{P_i[j]}{P'_i[j']} > \delta, i < i'\}$$
(6)

$$basisPatternConnectivity(\{a_{ki}|1 \le i \le n\}) \qquad \forall \ 1 \le k \le K$$
 (7)

Drawback:

Does not scale to full-scale realistic instances; poor propagation of experimental noise.





SAT Modulo Theory (SMT) Approach: Experimental Validation

Al-Li-Fe instance with 6 phases:

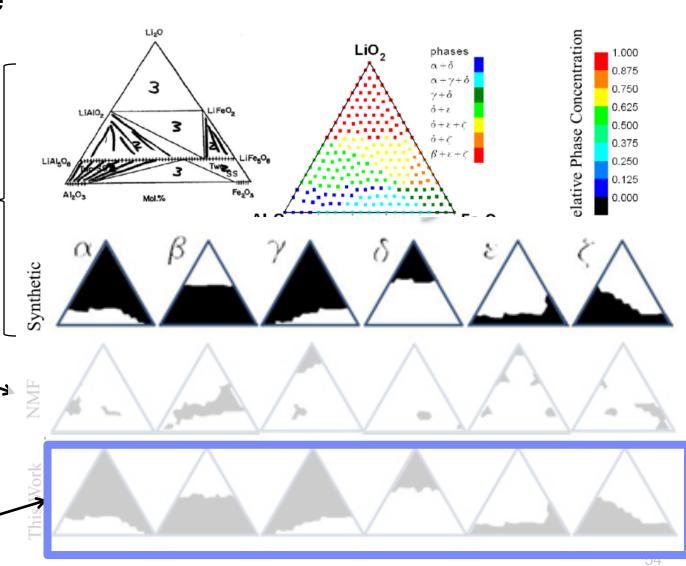
Ground truth (known)

Previous work

(ML Based approaches) violates many physical requirements

Our SMT

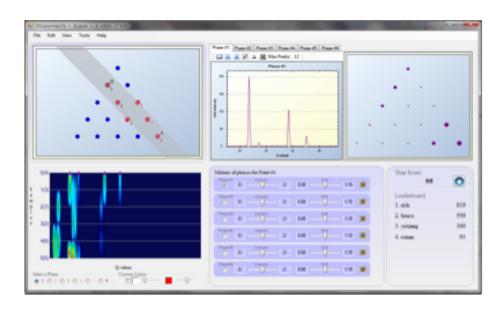
approach is much more robust

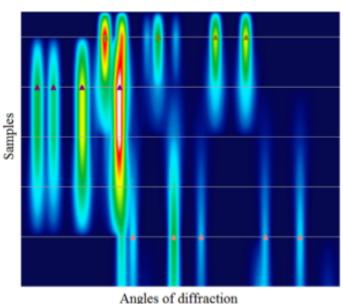






- Can human input boost the performance of combinatorial reasoning and optimization methods (SMT)?
- Human input about potential grouping of diffraction peaks collected on Amazon's Mechanical Turk. (About 500 patterns)
 Provides info on potential so-called backdoor variables.







		Data	set			Time w/o	Time w/	# assigned
System	P	L^*	K	#var	#cst	user input (s)	user input (s)	var. by user
A/B/C	36	8	4	408	2095	3502	150	19 (4.6%)
A/B/C	60	8	4	624	3369	17345	261	18 (2.9%)
Al/Li/Fe	15	6	6	267	1009	79	27	6 (2.2%)
Al/Li/Fe	28	6	6	436	1864	346	83	12 (2.7%)
Al/Li/Fe	28	8	6	490	2131	10076	435	26 (5.3%)
Al/Li/Fe	28	10	6	526	2309	28170	188	23 (4.3%)
Al/Li/Fe	45	7	6	693	3281	18882	105	28 (4.0%)
Al/Li/Fe	45	8	6	711	3410	46816	74	30 (4.2%)

Human input can dramatically speed up the performance of combinatorial optimization methods (identification of backdoor variables).

Leverages the complementary strength of human input, providing global insights into problem structure, and the power of combinatorial solvers to exploit complex constraints.

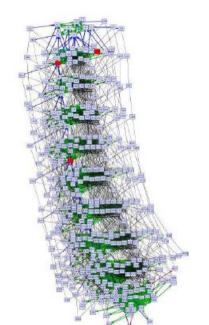
Note: Human input identifies subsets of peaks that (quite likely) go together.

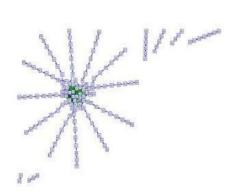
Observation

Visual representation allows human input to be effective in discovering structure and setting backdoor variables in this domain.

Contrast with earlier work:

SAT encodings of planning problems. It was very difficult to give any semantic interpretation to the backdoor variables of the problem instances. Synthetic planning formulas with log(n) size backdoors (provably).





Setting 2 out of 392 variables.

(Hoffmann, Gomes, Selman AIPS-06)



Conclusions

We have discussed how human insights can be combined with reasoning and optimization methods for scientific discovery.

Finite Mathematics --- Obtained first procedure for Spatially Balanced Latin Square construction.

Materials Discovery --- Boosted interpretation of X-ray diffraction patterns with crowdsourced input.