

MATHCHECK2: A SAT+CAS Verifier for Combinatorial Conjectures

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Motivation

The research areas of SMT [SAT-Modulo-Theories] solving and symbolic computation are quite disconnected. On the one hand, SMT solving has its strength in efficient techniques for exploring Boolean structures, learning, combining solving techniques, and developing dedicated heuristics, but its current focus lies on easier theories and it makes use of symbolic computation results only in a rather naive way.

Erica Ábrahám¹

¹Building bridges between symbolic computation and satisfiability checking. *Proceedings of the 2015 International Symposium on Symbolic and Algebraic Computation.*

Satisfiability checking

Problem statement

Given a **logical formula**, determine if it is **satisfiable**.

- ▶ A **logical formula** is an expression involving Boolean variables and logical connectives such as \wedge , \vee , \neg .
- ▶ A formula is **satisfiable** if there exists an assignment to the variables which make the formula true.

Example

Is $(x \vee y \vee \neg z) \wedge (\neg x \vee \neg y) \wedge z$ satisfiable?

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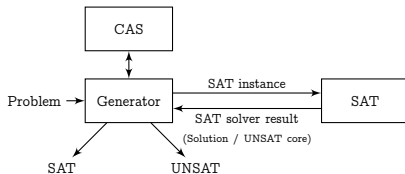
Example

Is $(x \vee y \vee \neg z) \wedge (\neg x \vee \neg y) \wedge z$ satisfiable?

Yes: Take x to be false, and take y and z to be true.

The MATHCHECK2 System

- ▶ Uses SAT and CAS functionality to finitely verify or counterexample conjectures in mathematics.
- ▶ Used to study conjectures in combinatorial design theory about the existence of Hadamard and Williamson matrices.



Authors

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Experimental Results

MATHCHECK2 was able to show that...

- ▶ Williamson matrices of order 35 do not exist.
 - ▶ Used under 9 hours of computation time on SHARCNET².
 - ▶ First shown by Đoković³, who requested an independent verification.
- ▶ Williamson matrices exist for all orders $n < 35$.
 - ▶ Even orders were mostly previously unstudied.
- ▶ Found over 160 Hadamard matrices which were not previously in the library of the CAS MAGMA.

²64-bit AMD Opteron processors running at 2.2 GHz

³Williamson matrices of order $4n$ for $n = 33, 35, 39$. *Discrete Mathematics*.

Hadamard matrices

- ▶ square matrix with ± 1 entries
- ▶ any two distinct rows are orthogonal

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Example

$$H = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ -1 & 1 & 1 & -1 \end{bmatrix}$$

Conjecture

An $n \times n$ Hadamard matrix exists for any n a multiple of 4.

Naive Hadamard Encoding

Each entry of H will be represented using a *Boolean variable* encoding with $BV(1) = \text{true}$ and $BV(-1) = \text{false}$.

Multiplication becomes XNOR under this encoding, i.e.,

$$BV(x \cdot y) = \neg(BV(x) \oplus BV(y)) \quad \text{for } x, y \in \{\pm 1\}.$$

Naive Hadamard Encoding

Arithmetic formula encoding

$$\sum_{k=0}^{n-1} h_{ik} \cdot h_{jk} = 0 \quad \text{for all } i \neq j.$$

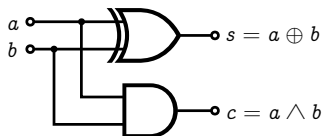
Boolean variable encoding

Using ‘product’ variables $p_{ijk} := \text{BV}(h_{ik} \cdot h_{jk})$ this becomes the cardinality constraints

$$\sum_{\substack{k=0 \\ p_{ijk} \text{ true}}}^{n-1} 1 = \frac{n}{2} \quad \text{for all } i \neq j.$$

Naive Hadamard Encoding

A *binary adder* consumes Boolean values and produces Boolean values; when thought of as bits, the outputs contain the binary representation of how many inputs were true.



To encode the cardinality constraints we use a network of binary adders with:

- ▶ n inputs (the variables $\{p_{ijk}\}_{k=0}^{n-1}$)
- ▶ $\lfloor \log_2 n \rfloor + 1$ outputs (counting the number of input variables which are true)

Williamson Matrices

- ▶ $n \times n$ matrices A, B, C, D
- ▶ entries ± 1
- ▶ symmetric, circulant
- ▶ $A^2 + B^2 + C^2 + D^2 = 4nI_n$

Symmetric and Circulant Matrices

Such matrices are defined by their first $\lceil \frac{n+1}{2} \rceil$ entries so we may refer to them as if they were sequences.

Examples ($n = 5$ and 6)

$$\begin{bmatrix} a_0 & a_1 & a_2 & a_2 & a_1 \\ a_1 & a_0 & a_1 & a_2 & a_2 \\ a_2 & a_1 & a_0 & a_1 & a_2 \\ a_2 & a_2 & a_1 & a_0 & a_1 \\ a_1 & a_2 & a_2 & a_1 & a_0 \end{bmatrix}$$

symmetric conditions

$$\begin{bmatrix} a_0 & a_1 & a_2 & a_3 & a_2 & a_1 \\ a_1 & a_0 & a_1 & a_2 & a_3 & a_2 \\ a_2 & a_1 & a_0 & a_1 & a_2 & a_3 \\ a_3 & a_2 & a_1 & a_0 & a_1 & a_2 \\ a_2 & a_3 & a_2 & a_1 & a_0 & a_1 \\ a_1 & a_2 & a_3 & a_2 & a_1 & a_0 \end{bmatrix}$$

circulant conditions

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symmetric conditions

$$\begin{bmatrix} a_0 & a_1 & a_2 & a_3 & a_2 & a_1 \\ a_1 & a_0 & a_1 & a_2 & a_3 & a_2 \\ a_2 & a_1 & a_0 & a_1 & a_2 & a_3 \\ a_3 & a_2 & a_1 & a_0 & a_1 & a_2 \\ a_2 & a_3 & a_2 & a_1 & a_0 & a_1 \\ a_1 & a_2 & a_3 & a_2 & a_1 & a_0 \end{bmatrix}$$

circulant conditions

Williamson Matrices Sequences

- ▶ sequences A, B, C, D of length $\lceil \frac{n+1}{2} \rceil$
- ▶ entries ± 1
- ▶ $\text{PAF}_A(s) + \text{PAF}_B(s) + \text{PAF}_C(s) + \text{PAF}_D(s) = 0$ for $s = 1, \dots, \lceil \frac{n-1}{2} \rceil$.

The PAF^4 here is defined

$$\text{PAF}_A(s) := \sum_{k=0}^{n-1} a_k a_{(k+s) \bmod n}.$$

⁴Periodic Autocorrelation Function

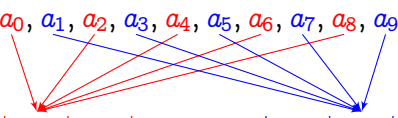
Compression

The m -compression of a sequence $A = [a_0, \dots, a_{n-1}]$ of length $n = dm$ is the sequence of length d

$$A^{(d)} := [a_0^{(d)}, \dots, a_{d-1}^{(d)}] \quad \text{where } a_j^{(d)} := \sum_{k=0}^{m-1} a_{j+kd}.$$

Example

The sequence $A = [a_0, a_1, a_2, a_3, a_4, a_5, a_6, a_7, a_8, a_9]$ has the 5-compression


$$A^{(2)} = [a_0 + a_2 + a_4 + a_6 + a_8, \quad a_1 + a_3 + a_5 + a_7 + a_9].$$

Compression

The m -compression of a sequence $A = [a_0, \dots, a_{n-1}]$ of length $n = dm$ is the sequence of length d

$$A^{(d)} := [a_0^{(d)}, \dots, a_{d-1}^{(d)}] \quad \text{where } a_j^{(d)} := \sum_{k=0}^{m-1} a_{j+kd}.$$

Example

The sequence $A = [a_0, a_1, a_2, a_3, a_4, a_5, a_6, a_7, a_8, a_9]$ has the 2-compression

$$A^{(5)} = [a_0 + a_5, a_1 + a_6, a_2 + a_7, a_3 + a_8, a_4 + a_9].$$

Useful Properties of Compressed Sequences

Lemma 1

The entries of an m -compression of a ± 1 -sequence of length dm :

- ▶ have absolute value at most m
- ▶ have the same parity as m

Lemma 2

The compression of a symmetric sequence is also symmetric.

Technique 1: Sum-of-squares Decomposition

Doković–Kotsireas⁵ theorem

Williamson sequences satisfy

$$\text{PAF}_{A^{(d)}}(0) + \text{PAF}_{B^{(d)}}(0) + \text{PAF}_{C^{(d)}}(0) + \text{PAF}_{D^{(d)}}(0) = 4n.$$

When $d = 1$, this becomes

$$\text{rowsum}(A)^2 + \text{rowsum}(B)^2 + \text{rowsum}(C)^2 + \text{rowsum}(D)^2 = 4n.$$

⁵Compression of periodic complementary sequences and applications.
Designs, Codes and Cryptography

Technique 1: Sum-of-squares Decomposition

Why is this useful?

CAS functions exist which can determine all possible solutions of

$$w^2 + x^2 + y^2 + z^2 = 4n \quad w, x, y, z \equiv n \pmod{2}.$$

This tells us all possibilities for the rowsums of A, B, C, D .

We can then use binary adders to encode the constraints

$$\begin{array}{ll} \text{rowsum}(A) = w & \text{rowsum}(B) = x \\ \text{rowsum}(C) = y & \text{rowsum}(D) = z \end{array}$$

and pass that information to the SAT solver.

Technique 1: Sum-of-squares Decomposition

Example

When $n = 35$, there are exactly three ways to write $4n$ as a sum of four positive odd squares in ascending order:

$$1^2 + 3^2 + 3^2 + 11^2 = 4 \cdot 35$$

$$1^2 + 3^2 + 7^2 + 9^2 = 4 \cdot 35$$

$$3^2 + 5^2 + 5^2 + 9^2 = 4 \cdot 35$$

Williamson Equivalences

Williamson sequences A, B, C, D can be re-ordered and negated without affecting the Williamson conditions.

Given this, we may enforce the constraint

$$0 \leq \text{rowsum}(A) \leq \text{rowsum}(B) \leq \text{rowsum}(C) \leq \text{rowsum}(D).$$

Technique 2: Divide-and-conquer

For efficiency reasons, we want to partition the search space into subspaces. An effective way to do this is to have each subspace contain one possibility for the compressions of A , B , C , D .

- ▶ Determine all possible compressions with Lemmas 1 and 2.
- ▶ Use the DK theorem to remove possibilities which are necessarily invalid (for example, because their *power spectral density* is too large).

Power Spectral Density

The *power spectral density* of a sequence A is

$$\text{PSD}_A(s) := |\text{DFT}_A(s)|^2$$

where DFT_A is the *discrete Fourier transform* of A .

Example

The power spectral density of $A = [1, 1, -1, -1, 1]$ is given by:

$$\begin{aligned}\text{PSD}_A(0) &= 1^2 &&= 1 \\ \text{PSD}_A(1) &\approx 3.236^2 &&= 10.472 \\ \text{PSD}_A(2) &\approx (-1.236)^2 &&= 1.528 \\ \text{PSD}_A(3) &\approx (-1.236)^2 &&= 1.528 \\ \text{PSD}_A(4) &\approx 3.236^2 &&= 10.472\end{aligned}$$

Đoković–Kotsireas Theorem

For all $s \in \mathbb{Z}$, Williamson sequences satisfy

$$\text{PSD}_A(s) + \text{PSD}_B(s) + \text{PSD}_C(s) + \text{PSD}_D(s) = 4n$$

and these **still hold** if A, B, C, D are replaced with their compressions.

Corollary

$\text{PSD}_X(s) \leq 4n$ if X is a Williamson sequence or a compression of a Williamson sequence.

Technique 2: Divide-and-conquer

For $n = 35$ with 7-compression, the following is one of 119 compressions which satisfy the DK conditions:

$$A^{(5)} = [5, 1, -3, -3, 1]$$

$$B^{(5)} = [-3, 3, -3, -3, 3]$$

$$C^{(5)} = [-3, 1, -1, -1, 1]$$

$$D^{(5)} = [1, -3, -3, -3, -3]$$

Technique 2: Divide-and-conquer

If n has more than one nontrivial factor it is possible to perform compression by both factors. This increases the number of subspaces, but decreases the size of each subspace.

Example

Using 5 and 7-compression on $n = 35$ lead to the following number of subspaces for each decomposition type:

Instance type	# subspaces
$1^2 + 3^2 + 3^2 + 11^2$	6960
$1^2 + 3^2 + 7^2 + 9^2$	8424
$3^2 + 5^2 + 5^2 + 9^2$	6290

Technique 3: UNSAT Core

When an instance is found to be unsatisfiable, some SAT solvers can generate an *UNSAT core* containing which of those variables lead to the UNSAT result. We can prune instances which **set the same variables to the same values**.

Example

The $n = 35$ instances contained 3376 variables but only 168 were set differently between instances (those which encode the rowsum and compression values).

Experimental Results

Timings on SHARCNET for Williamson orders $25 \leq n \leq 35$ are below. The number of SAT calls which successfully returned a result is in parenthesis. A hyphen denotes a timeout after 24h.

Order	Base	Sum-of-squares	Divide-and-conquer	UNSAT Core
25	317s (1)	1702s (4)	408s (179)	408s (179)
26	865s (1)	3818s (3)	61s (3136)	34s (1592)
27	5340s (1)	8593s (3)	1518s (14994)	1439s (689)
28	7674s (1)	2104s (2)	234s (13360)	158s (439)
29	-	21304s (1)	N/A	N/A
30	1684s (1)	36804s (1)	139s (370)	139s (370)
31	-	83010s (1)	N/A	N/A
32	-	-	96011s (13824)	95891s (348)
33	-	-	693s (8724)	683s (7603)
34	-	-	854s (732)	854s (732)
35	-	-	31816s (21674)	31792s (19356)

Conclusion

We have...

- ▶ Outlined how MATHCHECK2 uses the power of SAT solvers in combination with domain specific knowledge and algorithms provided by computer algebra systems.
- ▶ Performed a requested verification of a nonexistence result using a new algorithm and techniques which generalize to other conjectures.
- ▶ Submitted new matrices to MAGMA's Hadamard database, including some generated by Williamson matrices of even order.