

The Pindakaas library

Encoding pseudo-Boolean and linear constraints to SAT

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The logo of KU Leuven, featuring the text "KU LEUVEN" in white, bold, uppercase letters on a dark blue rectangular background.

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The **unreasonable** effectiveness of **SAT solver** reasoning

SAT solvers can solve complex and hard SAT problems

- MiniZinc Challenge'24
 - Aircraft Disassembly Scheduling
 - Community Detection
- XCSP3 competition'24
- Applications: see SAT handbook ¹

Rank	Solver	Score
#1	cp-sat	1,557.75
#2	picat-sat	1,479.17
#3	chuffed	1,478.11
#4	gurobi	1,286.07
#5	cplex	1,258.59
#6	izplus	1,159.11
#7	cp-optimizer	1,137.28
#8	choco-solver_cp	1,128.72

Tab. 1: MiniZinc Challenge 2024

¹A. Biere, M. J. H. Heule, H. van Maaren, and T. Walsh, Eds. [1]

SAT solvers are also **versatile**

- Incremental solving
- Verifiable solving via proof logging
- Other variants e.g. optimization (via MaxSAT)
- Model counting
- User propagators

Highlight: automated theorem proving

- E.g. of the chromatic packing problem ²
- Combines many of these techniques to produce a 122TB proof closing the problem

²B. Subercaseaux and M. J. H. Heule [2]

³<https://bsubercaseaux.github.io/blog/2023/packingchromatic/>

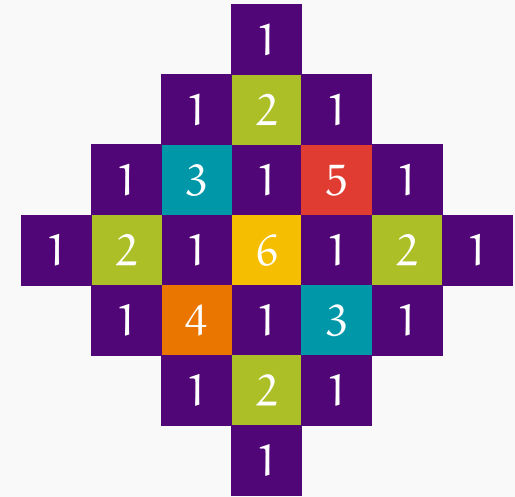
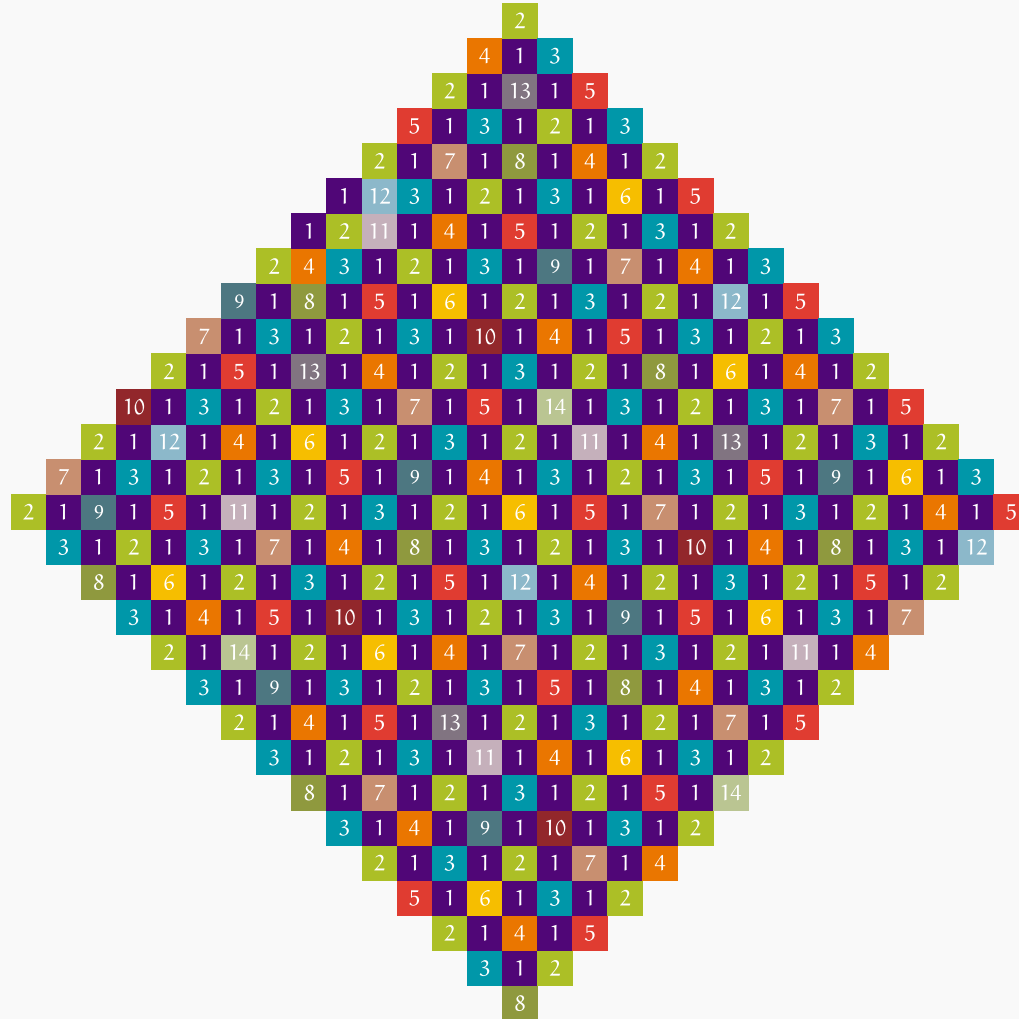


Fig. 1: Colouring a grid so that colour i is at least i steps away from the closest i .³

Should we checkerboard the grid?



None of these problems are SAT problems

Pindakaas: solve high-level problems with low-level solvers



- SAT solvers are efficient solvers of the **low-level** SAT problem
- But they can solve **high-level** problems only once translated (i.e. **encoded**)
- The quality of the encoding is **crucial** for solver performance

Pindakaas is a library which provides effective encoding methods and convenient solver access

Pindakaas: features and integrations

- Features
 - The encoding of high-level constraints to SAT:
 - Propositional
 - Pseudo-Boolean
 - Integer linear
 - Efficient and open-source Rust implementation + Python interface
 - Three SOTA, incremental back-end SAT solvers
- Integrations
 - CPMpy** A Constraint Programming and Modeling library @ KU Leuven
 - Huub** A Lazy Clause Generation solver developed @ Monash University

The SAT problem in Conjunctive Normal Form (CNF)

- A conjunction (i.e. AND, $\wedge$)
 - of disjunctions (i.e., ORs, **clauses**, $\vee$)
 - of literals:
 - variable, x
 - negated variable, $\neg x$

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Encoding and solving the **XOR-problem** for Boolean variables x, y :

$$x \oplus y$$

$$(x \vee y) \wedge (\neg x \vee \neg y) \quad \text{dec. } x$$

$$\neg y \implies \neg y$$

$$\top \quad \text{fixp.}$$

$$\{(x \mapsto \text{true}), (y \mapsto \text{false})\}$$

Usage from CPMpy

```
import cpmPy as cp
x = cp.boolvar()
y = cp.boolvar()
model = cp.Model(cp.Xor([x, y]))
assert model.solve(solver="pindakaas") is True
assert x.value() != y.value()
```

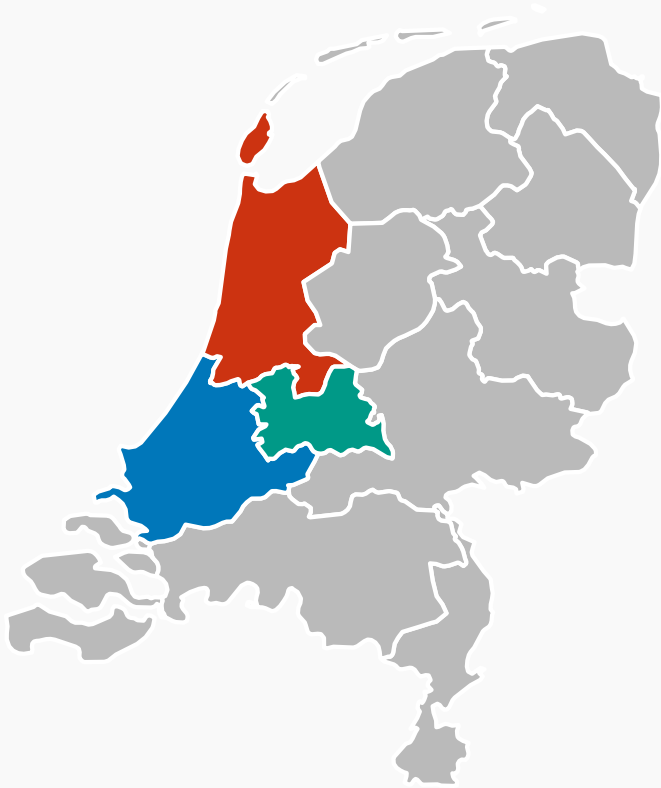
Advantage: many more supported constraint types and other solvers

Native usage through Python

```
import pindakaas as pdk
cadical = pdk.CaDiCaL()
x, y = cadical.new_vars(2)
cadical += x ^ y
with cadical.solve() as result:
    assert result.status == pdk.solver.Status.SATISFIED
    assert result.value(x) != result.value(y)
```

Advantage: wider scope, more features (e.g. more control over encoding)

Encoding a n region, k colour map-colouring problem



Let **integer** variable x denote the colour assigned to North-Holland, y the colour to South-Holland, z to Utrecht, ...

$$x \in [1..k] \wedge y \in [1..k] \wedge z \in [1..k] \wedge \dots \\ \wedge x \neq y \wedge x \neq z \wedge \dots$$

A partial assignment:

$$\{(x \mapsto 1), (y \mapsto 2), (z \mapsto 3)\}$$

Fig. 3: Map colouring
where $n = 12, k = 3$

The **direct** encoding

- Encode variable x using k Boolean variables $\llbracket x = 1 \rrbracket, \dots, \llbracket x = k \rrbracket$:

	$\llbracket x = 1 \rrbracket$	$\llbracket x = 2 \rrbracket$	$\llbracket x = 3 \rrbracket$
$(x \mapsto 1)$	true	false	false
$(x \mapsto 2)$	false	true	false
$(x \mapsto 3)$	false	false	true

- Encode **domain constraint** $x \in [1..k]$
 - To prevent e.g. $\{(\llbracket x = 1 \rrbracket \mapsto \text{true}), (\llbracket x = 2 \rrbracket \mapsto \text{true})\}$

$$\llbracket x = 1 \rrbracket + \dots + \llbracket x = k \rrbracket = 1$$

- Encode **disequality constraint** on adjacent nodes, e.g. $x \neq y$:

$$\llbracket x = 1 \rrbracket \oplus \llbracket y = 1 \rrbracket \wedge \dots \wedge \llbracket x = k \rrbracket \oplus \llbracket y = k \rrbracket$$

Map colouring via CPMpy-pindakaas

```
import cpmPy as cp
# Create 12 integer variables with domains `[1..k]`
xs = cp.intvar(1, 3, shape=12)
# Create model with adjacency constraints
model = cp.Model(
    [
        xs[0] != xs[1],
        xs[0] != xs[2],
        # [...]
    ]
)
model.solve(solver="pindakaas")
```

Encoding with partial assignment $\{(x \mapsto 1), (y \mapsto 2), (z \mapsto 3)\}$

[illegible]

What is a good encoding?

- This encoding is **large** but propagates efficiently
 - The solver infers this partial assignment will **not** lead to a solution **without** search decisions
- Other encodings are small yet propagate poorly
 - Trade-off between **encoding size** and **propagation strength**
- These and other complex choices arise during encoding

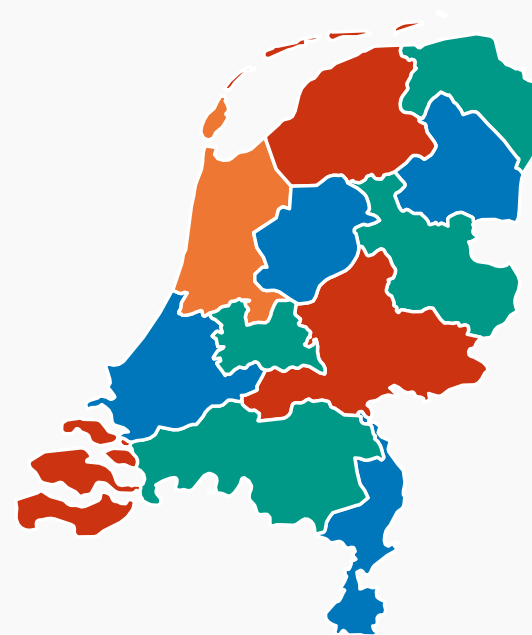


Fig. 4: Solution for $k = 4$

Single-constant multiplication

Encoding a integer linear constraint for integer variables, x_i :

$$\sum c_i x_i \leq c$$

- We can encode x_i with a **binary** representation
- But then we need to multiply by a constant: $c_i x_i = y_i$
- Fortunately, well-studied in the **hardware circuit design** community, so:

[CPAIOR'24] ⁴ Do the benefits carry over from hardware to software?

⁴H. Bierlee, J. J. Dekker, V. Lagoon, P. J. Stuckey, and G. Tack [3]

Finding optimal SCM encodings

- We decompose the multiplication into **additions**, **subtractions**, and **multiplications by 2^s** (i.e. left-shift by $s \geq 0$)
 - E.g. for $5x$, we encode $x + 4x = 5x$
 - E.g. for $11x$, we encode $x + 8x = 9x \wedge 9x + 2x = 11x$
- Why? Because additions and subtractions are cheap(er) and shifts are free
 - But minimizing number of additions/subtractions is NP-hard ⁵
 - We cannot compute the optimal decomposition at encode time

Instead, **pre-compute** $c \cdot x$ for $1 \leq c \leq 2047$

⁵P. Cappello and K. Steiglitz [4]

Static results (i.e. encoding sizes)

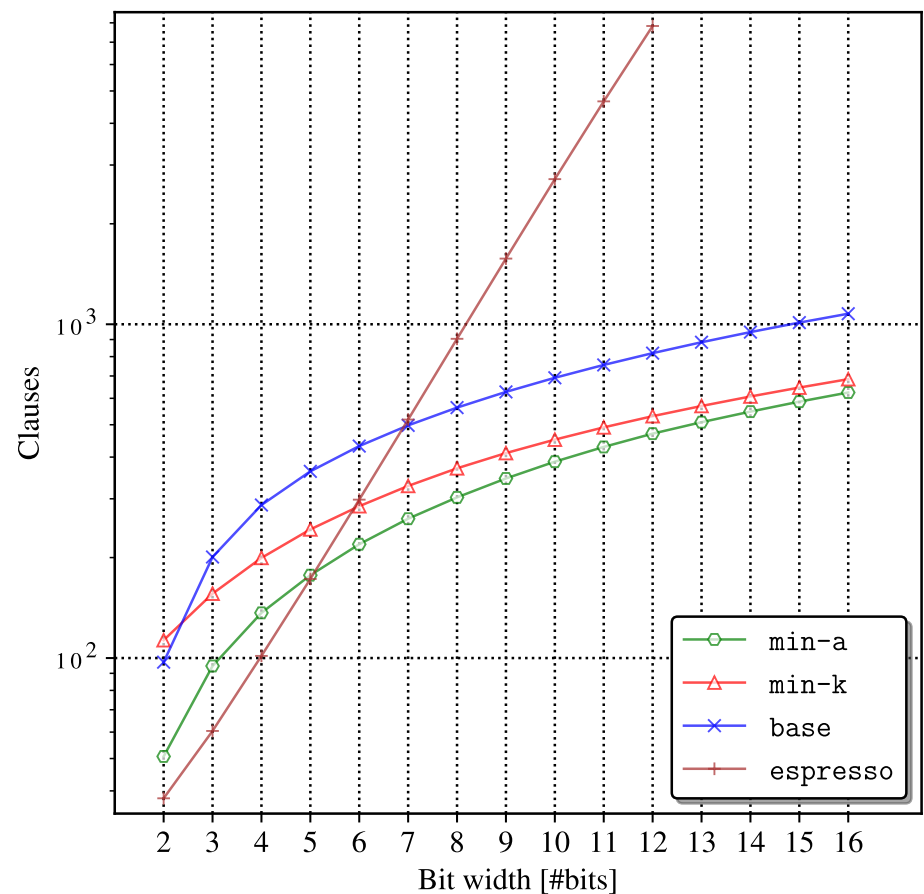
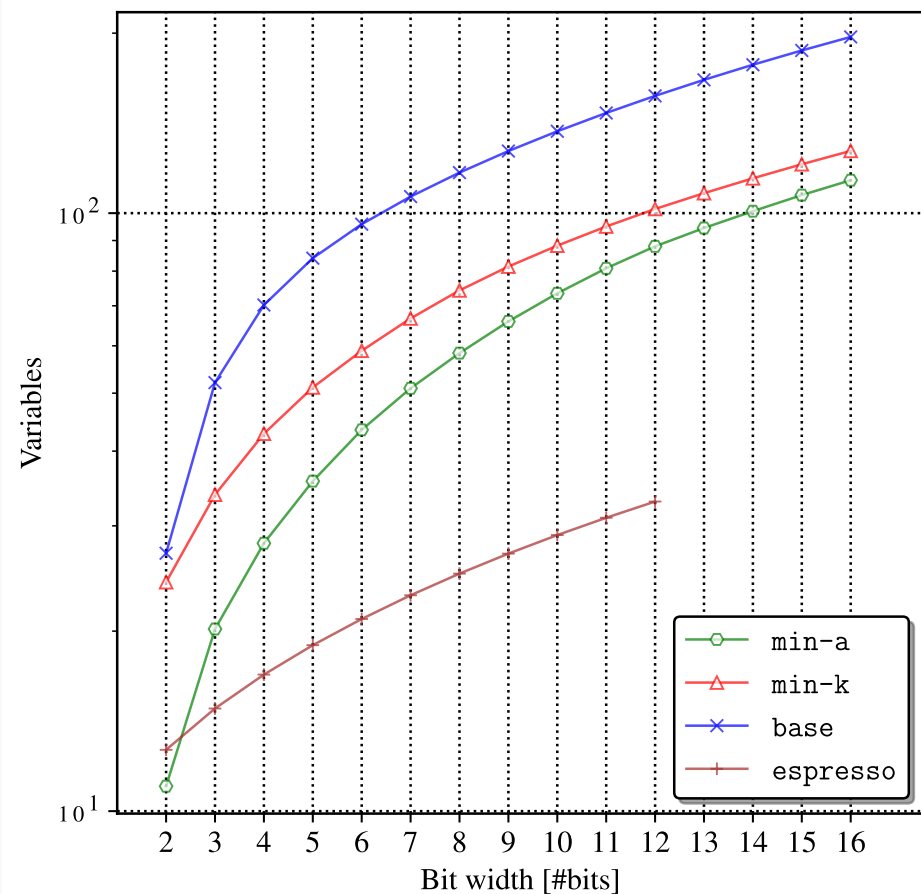


Fig. 5: Avg. #variables/#clauses for $c \cdot x, 1 \leq c \leq 2047, 2 \leq \text{bits}(x) \leq 16$

Challenges and future research

Challenges

- Ongoing challenge to release advanced features
 - Learning to cut features rather than stabilize them (now)
- Stiff competition from RustSAT (but competition is good)
 - Supporting a wider range of traditional encodings and features
 - But we focus more on higher-level constraints

Future research

- Proof-logging to certify the encoding is correct
- Inclusion in next version of CPMpy to run on wider sets of benchmarks
- Open to discuss new features to support collaborations!

Thank you for listening

Questions?



<https://github.com/pindakaashq/pindakaas>

References

- [1] A. Biere, M. J. H. Heule, H. van Maaren, and T. Walsh, Eds., *Handbook of Satisfiability*, 2nd ed. IOS Press, 2021.
- [2] B. Subercaseaux and M. J. H. Heule, “The Packing Chromatic Number of the Infinite Square Grid Is 15,” in *Tools and Algorithms for the Construction and Analysis of Systems - 29th International Conference, TACAS 2023, Held as Part of the European Joint Conferences on Theory and Practice of Software, ETAPS 2022, Paris, France, April 22-27, 2023, Proceedings, Part I*, S. Sankaranarayanan and N. Sharygina, Eds., in Lecture Notes in Computer Science, vol. 13993. Springer, 2023, pp. 389–406. doi: 10.1007/978-3-031-30823-9_20.
- [3] H. Bierlee, J. J. Dekker, V. Lagoon, P. J. Stuckey, and G. Tack, “Single Constant Multiplication for SAT,” in *Integration of Constraint Programming, Artificial Intelligence, and Operations Research - 21st International Conference, CPAIOR 2024, Uppsala, Sweden, May 28-31, 2024, Proceedings, Part I*, B. Dilkina, Ed., in Lecture Notes in Computer Science, vol. 14742. Springer, 2024, pp. 84–98. doi: 10.1007/978-3-031-60597-0_6.
- [4] P. Cappello and K. Steiglitz, “Some Complexity Issues in Digital Signal Processing,” *IEEE Transactions on Acoustics, Speech, and Signal Processing*, vol. 32, no. 5, pp. 1037–1041, 1984.