



The Picat-SAT Compiler

Neng-Fa Zhou
(CUNY Brooklyn College & GC)

Joint work with *Håkan* Kjellerstrand



Outline

- Why Picat-SAT?
- Picat's constraint modules
- Booleanization of constraints
 - Sign-and-magnitude log encoding of domain variables
 - Primitive, reified, and implicative constraints
 - The addition constraint $X+Y = Z$
 - The multiplication constraint $X*Y = Z$
 - Table constraints
 - Global constraints
- Experimental results
- Conclusion



Why Picat-SAT?

- SAT solvers' performance has drastically improved during the past 20 years
- SAT has become the backbone of many software systems
- It is viable to build SAT-based CSP solvers
- There are so many papers but so few compilers
 - Sugar, FznTini, BEE, and meSAT
- Build a SAT compiler that is solid, efficient, and practically usable



Picat's Constraint Modules

```
import cp.
```

```
import sat.
```

```
import mip.
```

■ Constraints

□ Domain

- $X :: \text{Domain}, X \text{notin Domain}$

□ Arithmetic

- $(X \# = Y), (X \# \neq Y), (X \# > Y), (X \# \geq Y), \dots$

□ Boolean

- $(X \# \setminus Y), (X \# \setminus / Y), (X \# \Leftrightarrow Y), (X \# \Rightarrow Y), (X \# \wedge Y), (\# \sim X),$

□ Table

- `table_in(VarTuple,Tuples), table_notin(VarTuple,Tuples)`

□ Global

- `all_different(L), element(I,L,V), circuit(L), cumulative(...), ...`

■ Solver invocation: `solve(Options,Vars)`



Picat's Constraint Modules

SEND + MORE = MONEY

```
import sat.

go =>
    Vars=[S,E,N,D,M,O,R,Y], % generate variables
    Vars :: 0..9,           % define the domains
    all_different(Vars),    % generate constraints
    S #!= 0,
    M #!= 0,
    1000*S+100*E+10*N+D+1000*M+100*O+10*R+E
        #= 10000*M+1000*O+100*N+10*E+Y,
    solve(Vars),           % search
    writeln(Vars).
```



Picat's Constraint Modules

Sudoku

```
import sat.
```

```
sudoku(Board) =>
```

```
  N = Board.len,
```

```
  Vars = Board.vars(),
```

```
  Vars :: 1..N,
```

```
  foreach (Row in Board) all_different(Row) end,
```

```
  foreach (J in 1..N)
```

```
    all_different([Board[I,J] : I in 1..N])
```

```
  end,
```

```
  M = round(sqrt(N)),
```

```
  foreach (I in 1..M..N-M, J in 1..M..N-M)
```

```
    all_different([Board[I+K,J+L] : K in 0..M-1, L in 0..M-1])
```

```
  end,
```

```
  solve(Vars).
```

```
Board = {{5,3,_,_,7,_,_,_,_},
         {6,_,_,1,9,5,_,_,_},
         {_,9,8,_,_,_,6,_,_},
         {8,_,_,6,_,_,3},
         {4,_,_,8,_,3,_,1},
         {7,_,_,2,_,_,6},
         {_,_,_,_,_,_,_,_},
         {_,_,_,_,_,_,_,_},
         {_,_,_,_,_,_,_,_},
         {_,_,_,_,_,_,_,_}}.
```



Booleanization of Constraints

- Domain variables
- Primitive, reified, and implicative constraints
- The addition constraint $X+Y = Z$
- The multiplication constraint $X*Y = Z$
- Table constraints
- Global constraints



Encoding Domain Variables

- Sparse encoding
 - $B \Leftrightarrow X = a$
One Boolean variable is used for each value
- Order encoding [Sugar, BEE, meSAT]
 - $B \Leftrightarrow X = a$
- Log encoding [FznTini]
 - $V: [a_1, \dots, a_n] \Leftrightarrow \langle B_m, \dots, B_1, B_0 \rangle$
Each domain variable is encoded as a binary string
 $m = \log_2(\max(|a_1|, |a_n|))$



Log-encoding

- Pros and Cons

- Pros: compact, and good for compiling arithmetic constraints
- Cons: less propagation power

- Encoding domains that have negative values

- 2's Complement Representation [FznTini]
 - Only one representation for 0
- Sign-and-Magnitude Representation [PicatSAT]
 - Efficient codes for $Y = \text{abs}(X)$ and $Y = -X$
 - One extra clause for banning negative zero

Encoding Domain Variables

$$X ::= [-2, -1, 1, 2] \Leftrightarrow \langle S, X_1, X_0 \rangle$$

■ A Naïve Code

$$\begin{aligned} S \vee X_1 \vee X_0 \\ \neg S \vee X_1 \vee X_0 \\ S \vee \neg X_1 \vee \neg X_0 \\ \neg S \vee \neg X_1 \vee \neg X_0 \end{aligned}$$

■ An Optimized Code (Using Espresso)

$$\begin{aligned} X_0 \vee X_1 \\ \neg X_0 \vee \neg X_1 \end{aligned}$$



Primitive, Reified, and Implicative Constraints

■ Primitive constraints

- $\sum B_i \geq c$ (at-least), $\sum B_i \leq c$ (at-most)
- $X = Y$, $X \geq Y$, $X > Y$, $X \neq Y$
- $-X = Y$, $\text{abs}(X) = Y$
- $X+Y = Z$, $X*Y = Z$

■ Reified and implicative constraints

- $B \Leftrightarrow C$ ($B \Rightarrow C$ and $\text{not}(C) \Rightarrow \text{not}(B)$)
- $B \Rightarrow C$

The Addition Constraint

$X+Y = Z$

- Unsigned addition

$$\begin{array}{r} X_{n-1} \dots X_1 X_0 \\ + Y_{n-1} \dots Y_1 Y_0 \\ \hline Z_n Z_{n-1} \dots Z_1 Z_0 \end{array}$$

Boolean variables are used for carries

- Signed addition

$$\begin{array}{l} X.s = 0 \wedge Y.s = 0 \Rightarrow Z.s = 0 \wedge X.m+Y.m = Z.m \\ X.s = 1 \wedge Y.s = 1 \Rightarrow Z.s = 1 \wedge X.m+Y.m = Z.m \\ X.s = 0 \wedge Y.s = 1 \wedge Z.s = 1 \Rightarrow X.m+Z.m = Y.m \\ X.s = 0 \wedge Y.s = 1 \wedge Z.s = 0 \Rightarrow Y.m+Z.m = X.m \\ X.s = 1 \wedge Y.s = 0 \wedge Z.s = 0 \Rightarrow X.m+Z.m = Y.m \\ X.s = 1 \wedge Y.s = 0 \wedge Z.s = 1 \Rightarrow Y.m+Z.m = X.m \end{array}$$

The Multiplication Constraint

$$X * Y = Z$$

- Shift-and-Add Algorithm (Pen and Pencil)

$$Y \Leftrightarrow \langle Y_{n-1}, \dots, Y_1, Y_0 \rangle$$

$$Y_0 = 0 \Rightarrow S_0 = 0$$

$$Y_0 = 1 \Rightarrow S_0 = X$$

$$Y_1 = 0 \Rightarrow S_1 = S_0$$

$$Y_1 = 1 \Rightarrow S_1 = (X \ll 1) + S_0$$

⋮

$$Y_{n-1} = 0 \Rightarrow S_{n-1} = S_{n-2}$$

$$Y_{n-1} = 1 \Rightarrow S_{n-1} = (X \ll (n-1)) + S_{n-2}$$

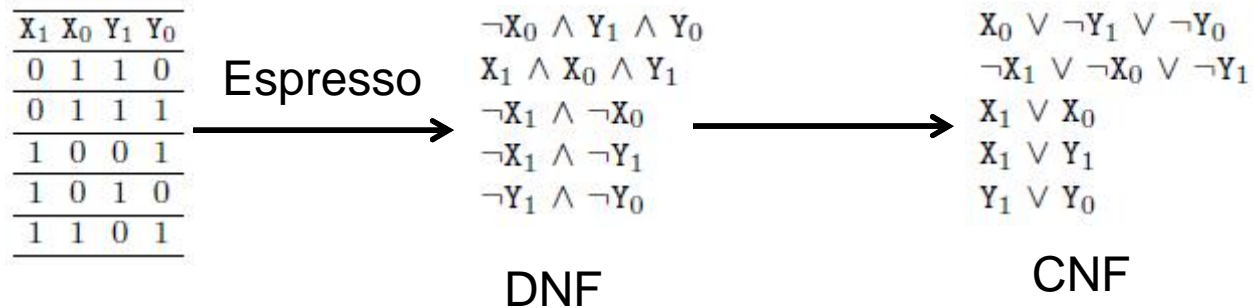
$$Z = S_{n-1}$$

Table Constraints

`table_in({X,Y},[{1,2}, {1,3}, {2,1}, {2,2}, {3,1}])`

$X \Leftrightarrow \langle X_1, X_0 \rangle$

$Y \Leftrightarrow \langle Y_1, Y_0 \rangle$





Global Constraints

- `all_different([V1, V2, ..., Vn])`

$V_i \neq V_j$ for $i, j = 1, \dots, n, i < j$

If it is an assignment-type constraint, then

Let a_1, a_2, \dots, a_n be the values.

$\sum (V_i = a_j) = 1$ for each a_j .

- `circuit([V1, V2, ..., Vn])`

`all_different([V1, V2, ..., Vn])`

and constraints for preventing sub-cycles



Global Constraints

- `cumulative(Ss, Ds, Rs, Limit)`
 - Time decomposition
 - Task decomposition
- `element(I, L, V)`
 - Decomposed to a table constraint if L is ground
 - Decomposed to Boolean constraints, otherwise



Implementation

- Specialization techniques
 - $X+c = Y, a*X = Y, \dots$
- Common sub-expression elimination
 - $X+Y+Z1 < b1, X+Y-Z2 > b2$
- The default SAT solver: lingeling 587f
- 5,000 lines of codes in Picat

Experimental Results (Picat-SAT vs. Sugar)

Benchmark	Picat			Sugar			Azucar		
	#vars	#cls	time	#vars	#cls	time	#vars	#cls	time
golombRuler-10-55	6438	27194	36.50	4584	190594	92.53	5635	44937	60.32
golombRuler-8	3024	12698	0.22	5111	422046	3.04	2988	35847	0.44
jss-ft10	13613	68880	5.90	76617	703313	6.24	8227	119126	1.50
knightTour-5	3051	21401	8.19	1296	20510	0.83	1667	9123	2.36
knightTour-7	11970	114818	374.36	4944	141818	99.68	5919	44059	85.53
magicSquare-5x5	2742	20302	1.68	3504	73717	0.72	2052	16239	0.29
magicSquare-9x9	33606	178328	48.58	60960	5463881	167.24	19405	317128	57.36
nqueens-200	1003112	3997099	1925.52	159196	18646098	131.96	284400	4004662	397.82
oss-gp03-01	785	3535	0.19	11126	46075	0.69	786	6231	2.13
oss-gp10-01	25540	139168	2.90	343189	6514888	40.40	17216	544880	5.79
socialGolfer-3-2-5	315	3360	0.12	810	2355	0.31	1680	5250	0.10
socialGolfer-6-3-7	9702	163863	13.19	22545	79236	13.78	46248	184980	14.04
tdsp-C1-1	6361	25187	0.67	1226	17631	0.26	4586	36997	0.64
tdsp-C2-1	16079	67032	5.09	2720	61572	0.72	11746	112436	7.12

Picat-SAT is competitive with Sugar on the Sugar benchmark suite overall, but is slower than Sugar on knightTour and nqueens because of the poor encoding of all_different .

Experimental Results (Picat-SAT vs. Sugar)

Benchmark	Picat			Sugar			Azucar		
	#vars	#cls	time	#vars	#cls	time	#vars	#cls	time
gcj_small	215	4746	0.275	n/a	n/a	n/a	n/a	n/a	n/a
gcj_large	980	6721	1.985	n/a	n/a	n/a	n/a	n/a	n/a
sendmory	370	2095	0.089	3775	88809	0.590	2501	85902	0.661
sendmory_neg	403	4082	0.144	3775	88809	0.608	4294	209632	1.730
crypta	3575	24452	1.353	n/a	n/a	n/a	443530	721220562	n/a
crypta_neg	3793	34676	3.054	n/a	n/a	n/a	430578	667692122	n/a

Picat-SAT significantly outperforms Sugar on arithmetic-intensive benchmarks.

Experimental Results

(Picat-SAT on Minizinc'15)

without global constraints

Benchmark	OR-Tools	Opturion CPX	iZplus	Picat-SAT
freepizza(MIN)	4.03	10.96	12.00	3.01
grid-colouring(MIN)	1.91	11.27	5.28	11.54
nmseq(SAT)	13.46	10.75	3.19	0.60
project-planning(MIN)	5.00	14.67	10.33	0.00
radiation(MIN)	0.00	11.04	8.56	10.40
triangular(MAX)	7.75	1.00	9.00	12.25
zephyrus(MIN)	9.00	7.50	7.50	6.00
(TOTAL)	41.16	67.19	55.86	43.79

with global constraints

Benchmark	OR-Tools	Opturion CPX	iZplus	Picat-SAT
costas-array(SAT)	7.35	4.86	5.06	0.73
cvrp(MIN)	7.07	8.59	12.30	2.04
gfd-schedule(MIN)	2.48	10.53	11.99	5.00
is(MIN) TOTAL	7.56	7.88	12.56	0.00
largescheduling(MIN)	13.00	10.00	0.00	0.00
mapping(MIN)	2.00	14.91	7.00	5.09
multi-knapsack(MAX)	4.18	10.75	4.26	9.80
opd(MIN)	8.07	2.50	7.00	12.43
open_stacks(MIN)	10.10	5.19	2.00	12.71
p1f(MIN)	14.53	3.95	2.01	3.51
roster(MIN)	11.72	4.44	11.40	2.44
spot5(MAX)	8.00	6.68	2.34	12.98
tdtsp(MIN)	8.24	11.26	10.50	0.00
(TOTAL)	104.31	101.54	88.43	66.72

Picat-CP vs. Picat-SAT on Minizinc'13

Benchmark	CP	SAT
black-hole	4	5
cargo	0	2
celar	5	4
filters	3	5
fjsp	3	5
ghoulomb	1	2
javarouting	3	5
l2p	1	5
league	2	5
mario	4	5
nmseq	0	2
nonogram	0	5
on-call-rostering	5	5
pattern-set-mining	5	1
pentominoes-int	4	0
proteindesign12	0	0
radiation	5	5
repsp	5	5
rubik	2	5
vrp	5	5
<i>Total</i>	62	76

- The settings and the result
 - An optimization instance is considered *solved* if at least one solution was found.
 - For CP, the labeling option *ffc* was used.
 - Total number instances: 100.
 - Time limit: 900s per instance.
 - SAT outperforms CP.



Summary and Future Work

■ Picat-SAT

- Sign-and-magnitude log encoding
- Uses some of the encoding algorithms and tools from hardware design (addition and multiplication, Espresso)
- Solid, efficient, and practically usable

■ Future Work

- More specialization and optimization techniques
- New encoding algorithms for PB and table constraints
- Efficient decomposition algorithms for global constraints