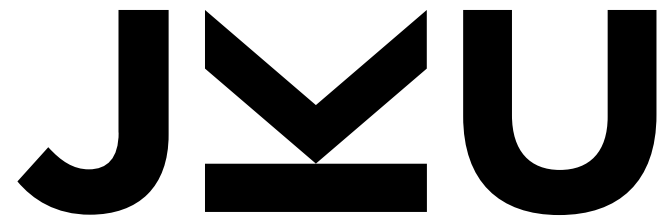


# Encoding into SAT

Armin Biere



**JOHANNES KEPLER  
UNIVERSITÄT LINZ**

**SAT/SMT/AR Summer School 2018**

Manchester, UK

July 3, 2018

# Dress Code Summer School Speaker as SAT Problem

- propositional logic:

- variables      **tie**    **shirt**
- negation       $\neg$             (not)
- disjunction     $\vee$             (or)
- conjunction     $\wedge$           (and)

- clauses (conditions / constraints)

1. clearly one should not wear a **tie** without a **shirt**

$$\neg \mathbf{tie} \vee \mathbf{shirt}$$

2. not wearing a **tie** nor a **shirt** is impolite

$$\mathbf{tie} \vee \mathbf{shirt}$$

3. wearing a **tie** and a **shirt** is overkill

$$\neg(\mathbf{tie} \wedge \mathbf{shirt}) \equiv \neg \mathbf{tie} \vee \neg \mathbf{shirt}$$

- Is this formula in conjunctive normal form (CNF) **satisfiable**?

$$(\neg \mathbf{tie} \vee \mathbf{shirt}) \wedge (\mathbf{tie} \vee \mathbf{shirt}) \wedge (\neg \mathbf{tie} \vee \neg \mathbf{shirt})$$

# SAT-Race 2010 Award

“Plingeling”  
by Armin Biere

is awarded the title of  
**Best Parallel Solver**

Chair of SAT-Race Organizing Committee

# SAT-Race 2010 Award

“Lingeling”  
by Armin Biere

is awarded the title of  
**Second Prize Winner**

Edinburgh, Scotland, July 14, 2010

Tenth International Conference  
On Theory and Applications of Satisfiability Testing

Competition 2007

**Gold medal**

Awarded to Ticocat 335 solver  
written by Armin Biere  
for best performance by a solver in the industrial category, satisfiability  
benchmarks specialty.

The SAT2007 competition judges and organizers

Twelfth International Conference  
On Theory and Applications of Satisfiability Testing

Competition 2009

**Gold medal**

Eighteenth International Conference  
on Theory and Applications of Satisfiability  
Testing

**SAT-Race 2011**

Parallel Track

**3rd Prize**

Awarded to  
Treengeling v1.5.0

Written by  
Armin Biere

Eighteenth International Conference  
on Theory and Applications of Satisfiability  
Testing

**SAT-Race 2015**

Main Track

**Special Prize:  
Most Innovative Solver**

Awarded to  
Lingeling v1.0.0

Written by  
Armin Biere

Twelfth International Conference  
On Theory and Applications of Satisfiability Testing

**SAT-Race 2010**

Parallel Track

**2nd Prize**

Awarded to  
Plingeling v1.0.0

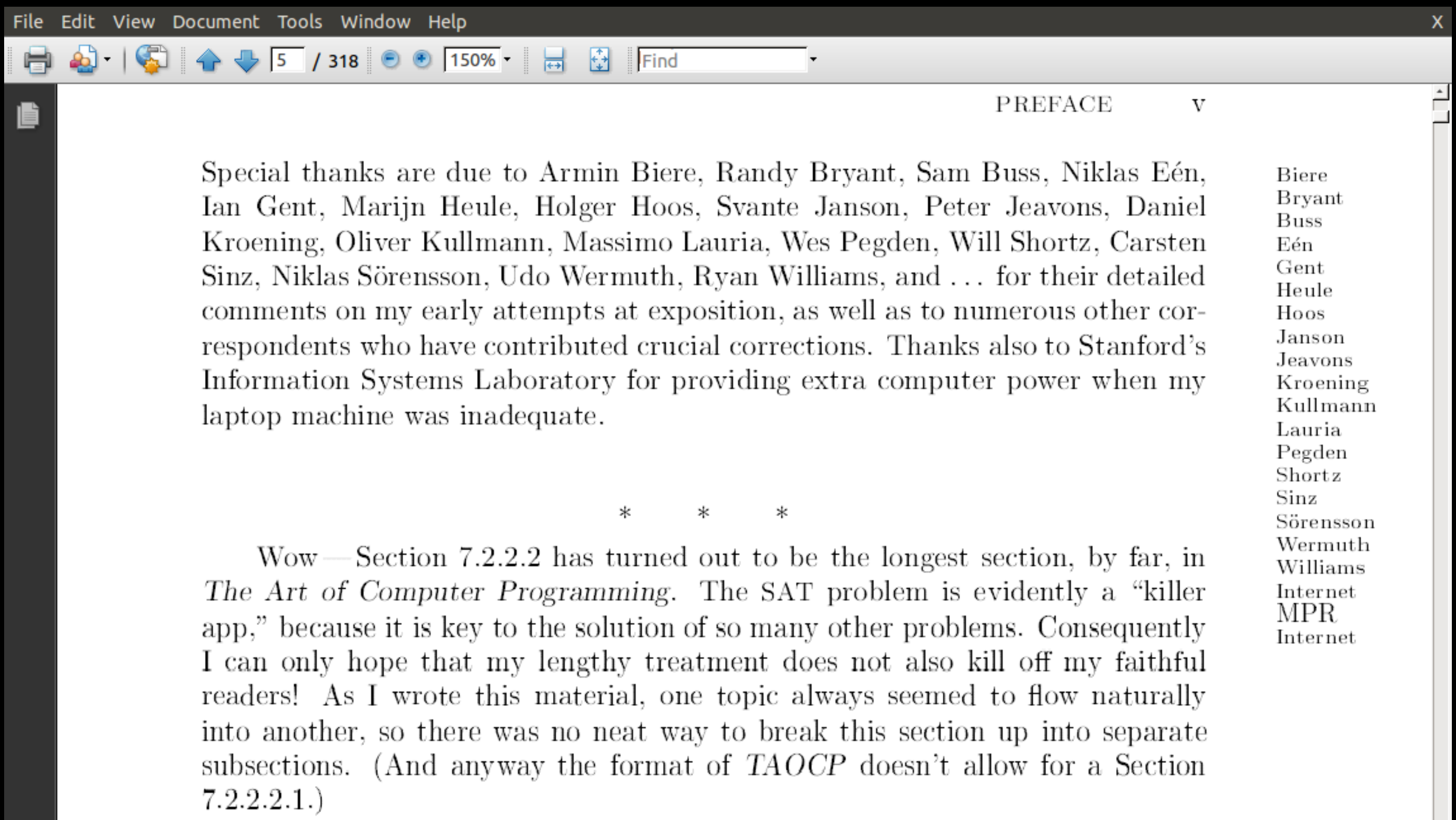
Written by  
Armin Biere











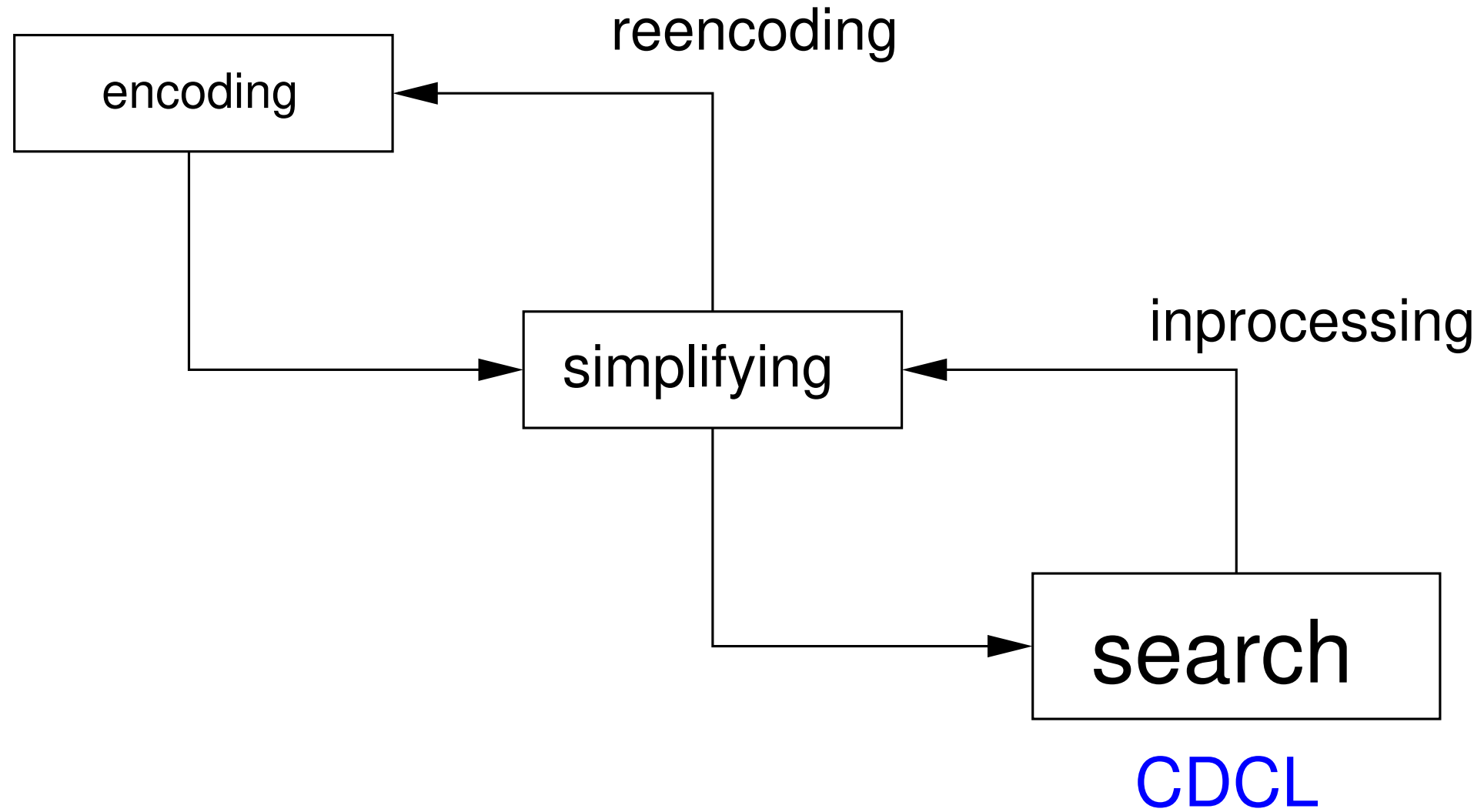
Special thanks are due to Armin Biere, Randy Bryant, Sam Buss, Niklas Eén, Ian Gent, Marijn Heule, Holger Hoos, Svante Janson, Peter Jeavons, Daniel Kroening, Oliver Kullmann, Massimo Lauria, Wes Pegden, Will Shortz, Carsten Sinz, Niklas Sörensson, Udo Wermuth, Ryan Williams, and . . . for their detailed comments on my early attempts at exposition, as well as to numerous other correspondents who have contributed crucial corrections. Thanks also to Stanford’s Information Systems Laboratory for providing extra computer power when my laptop machine was inadequate.

Biere  
Bryant  
Buss  
Eén  
Gent  
Heule  
Hoos  
Janson  
Jeavons  
Kroening  
Kullmann  
Lauria  
Pegden  
Shortz  
Sinz  
Sörensson  
Wermuth  
Williams  
Internet  
MPR  
Internet

\* \* \*

Wow—Section 7.2.2.2 has turned out to be the longest section, by far, in *The Art of Computer Programming*. The SAT problem is evidently a “killer app,” because it is key to the solution of so many other problems. Consequently I can only hope that my lengthy treatment does not also kill off my faithful readers! As I wrote this material, one topic always seemed to flow naturally into another, so there was no neat way to break this section up into separate subsections. (And anyway the format of *TAOCP* doesn’t allow for a Section 7.2.2.2.1.)

# What is Practical SAT Solving?





$$\text{Schur}(2) = 4 \ 5$$

see also: Marijn Heule, "Schur Number Five", AAI'18.

$\text{Schur}(k) = n$  iff

$n$  is the largest number such that

$N = \{1, \dots, n\}$  can be colored with  $k$  colors

and if  $i + j = k$  for  $i, j, k \in N$  then they do not have the same color

such equations are not monochromatic

$$1 + 1 = 2 \quad 2 + 2 = 4 \quad 1 + 3 = 4$$

$$1 + 4 = 5 \quad 2 + 3 = 5$$

encoded in SAT ( $x = 1$  iff  $x$  is colored with first color)

$$(\bar{x}_1 \vee \bar{x}_1 \vee \bar{x}_2) \wedge (x_1 \vee x_1 \vee x_2) \wedge (\bar{x}_2 \vee \bar{x}_2 \vee \bar{x}_4) \wedge (x_2 \vee x_2 \vee x_4) \wedge (\bar{x}_1 \vee \bar{x}_3 \vee \bar{x}_4) \wedge (x_1 \vee x_3 \vee x_4)$$

$$(\bar{x}_1 \vee \bar{x}_4 \vee \bar{x}_5) \wedge (x_1 \vee x_4 \vee x_5) \wedge (\bar{x}_2 \vee \bar{x}_3 \vee \bar{x}_5) \wedge (x_2 \vee x_3 \vee x_5)$$

# DIMACS Encoder for Schur(2)

$$1 + 3 = 4$$

```
#include <stdio.h>
#include <stdlib.h>
int main (int argc, char ** argv) {
    int n = argc > 1 ? atoi (argv[1]) : 5;
    printf ("p cnf %d 0\n", n); // FIXME
    for (int i = 1; i <= n; i++)
        for (int j = i; j <= n; j++) {
            int k = i + j;
            if (k <= n)
                printf ("%d %d %d 0\n", i, j, k),
                printf ("%d %d %d 0\n", -i, -j, -k);
        }
    return 0;
}
```

see also: Heule, Kullmann, Marek, "Solving and Verifying the boolean Pythagorean Triples problem via Cube-and-Conquer", SAT'16.

# DIMACS Encoder for Pythagorean Triples Problem

$$3^2 + 4^2 = 5^2$$

```
#include <math.h>
#include <stdio.h>
#include <stdlib.h>
int main (int argc, char ** argv) {
    int n = argc > 1 ? atoi (argv[1]) : 7825;
    printf ("p cnf %d 0\n", n); // FIXME
    for (int i = 1; i <= n; i++)
        for (int j = i; j <= n; j++) {
            int k = sqrt (i*i + j*j);
            if (k <= n && i*i + j*j == k*k)
                printf ("%d %d %d 0\n", i, j, k),
                printf ("%d %d %d 0\n", -i, -j, -k);
        }
    return 0;
}
```

# Equivalence Checking If-Then-Else Chains

## original C code

```
if(!a && !b) h();  
else if(!a) g();  
else f();
```

↓

```
if(!a) {  
    if(!b) h();  
    else g();  
} else f();
```

⇒

## optimized C code

```
if(a) f();  
else if(b) g();  
else h();
```

↑

```
if(a) f();  
else {  
    if(!b) h();  
    else g(); } }
```

How to check that these two versions are equivalent?

# Compilation

$$\begin{aligned} \textit{original} &\equiv \mathbf{if} \neg a \wedge \neg b \mathbf{ then } h \mathbf{ else if } \neg a \mathbf{ then } g \mathbf{ else } f \\ &\equiv (\neg a \wedge \neg b) \wedge h \vee \neg(\neg a \wedge \neg b) \wedge \mathbf{if} \neg a \mathbf{ then } g \mathbf{ else } f \\ &\equiv (\neg a \wedge \neg b) \wedge h \vee \neg(\neg a \wedge \neg b) \wedge (\neg a \wedge g \vee a \wedge f) \end{aligned}$$

$$\begin{aligned} \textit{optimized} &\equiv \mathbf{if} a \mathbf{ then } f \mathbf{ else if } b \mathbf{ then } g \mathbf{ else } h \\ &\equiv a \wedge f \vee \neg a \wedge \mathbf{if} b \mathbf{ then } g \mathbf{ else } h \\ &\equiv a \wedge f \vee \neg a \wedge (b \wedge g \vee \neg b \wedge h) \end{aligned}$$

$$(\neg a \wedge \neg b) \wedge h \vee \neg(\neg a \wedge \neg b) \wedge (\neg a \wedge g \vee a \wedge f) \not\equiv a \wedge f \vee \neg a \wedge (b \wedge g \vee \neg b \wedge h)$$

satisfying assignment gives counter-example to equivalence

# Negation Normal Form

**Assumption:** we only have conjunction, disjunction and negation as operators.

Formula is in Negation Normal Form (NNF),  
if negations only occur in front of variables

⇒ all internal nodes in the formula tree are either ANDs or ORs

linear algorithms for generating NNF from an arbitrary formula

NNF generations includes elimination of non-monotonic operators (XOR, XNOR)

NNF of  $f \leftrightarrow g$  is NNF of  $f \wedge g \vee \bar{f} \wedge \bar{g}$

in this case the result can be exponentially larger without sharing  
(parity, counting, adders, ...).

# NNF Algorithm

```
Formula formula2nnf (Formula f, Boole sign)
{
    if (is_variable (f))
        return sign ? new_not_node (f) : f;

    if (op (f) == AND || op (f) == OR)
    {
        l = formula2nnf (left_child (f), sign);
        r = formula2nnf (right_child (f), sign);
        flipped_op = (op (f) == AND) ? OR : AND;
        return new_node (sign ? flipped_op : op (f), l, r);
    }
    else
    {
        assert (op (f) == NOT);
        return formula2nnf (child (f), !sign);
    }
}
```

# Simple Translation of Formula into CNF

```
Formula formula2cnf (Formula f)
{
    if (is_cnf (f)) return f;

    if (op (f) == AND)
    {
        l = formula2cnf (left_child (f));
        r = formula2cnf (right_child (f));
        return new_node (AND, l, r);
    }
    else
    {
        assert (op (f) == OR);
        l = formula2cnf (left_child (f));
        r = formula2cnf (right_child (f));
        return merge_cnf (l, r);
    }
}
```



# Merging two CNFs

```
Formula merge_cnf (Formula f, Formula g)
{
  res = new_constant_node (TRUE);
  for (c = first_clause (f); c; c = next_clause (f, c))
    for (d = first_clause (g); d; d = next_clause (g, d))
      res = new_node (AND, res, new_node (OR, c, d));
  return res;
}
```

$$f \vee g \equiv \bigwedge_{i=1}^m C_i \vee \bigwedge_{j=1}^n D_j \equiv \bigwedge_{i=1}^m \bigwedge_{j=1}^n (C_i \vee D_j)$$

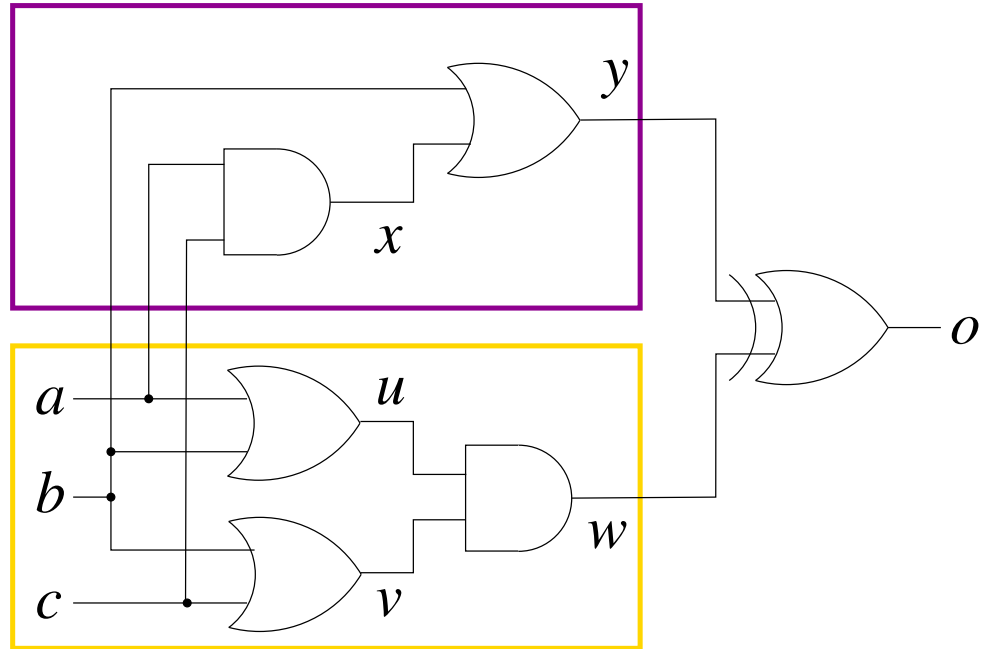
# Formula to NNF to CNF

```
Formula encode (Formula f)
{
    Formula nnf = formula2nnf (f, 0); // cheap
    Formula cnf = formula2cnf (nnf); // expensive
    return cnf;
}
```

- NNF translation is linear (even for circuits)
- one merge operation is already quadratic  $\Rightarrow$  NNF to CNF exponential
- whole “Formula to NFF to CNF” flow exponential
  - exponential in the number of alternations between OR and AND
- sometimes compact:  $(a \wedge c) \vee b \equiv (a \vee b) \wedge (c \vee b)$

preprocessing!

# Tseitin Transformation: Circuit to CNF



$$\begin{aligned}
 & o \wedge \\
 & (x \leftrightarrow a \wedge c) \wedge \\
 & (y \leftrightarrow b \vee x) \wedge \\
 & (u \leftrightarrow a \vee b) \wedge \\
 & (v \leftrightarrow b \vee c) \wedge \\
 & (w \leftrightarrow u \wedge v) \wedge \\
 & (o \leftrightarrow y \oplus w)
 \end{aligned}$$

$$o \wedge (x \rightarrow a) \wedge (x \rightarrow c) \wedge (x \leftarrow a \wedge c) \wedge \dots$$

$$o \wedge (\bar{x} \vee a) \wedge (\bar{x} \vee c) \wedge (x \vee \bar{a} \vee \bar{c}) \wedge \dots$$

# Tseitin Transformation: Gate Constraints

Negation:  $x \leftrightarrow \bar{y} \Leftrightarrow (x \rightarrow \bar{y}) \wedge (\bar{y} \rightarrow x)$   
 $\Leftrightarrow (\bar{x} \vee \bar{y}) \wedge (y \vee x)$

Disjunction:  $x \leftrightarrow (y \vee z) \Leftrightarrow (y \rightarrow x) \wedge (z \rightarrow x) \wedge (x \rightarrow (y \vee z))$   
 $\Leftrightarrow (\bar{y} \vee x) \wedge (\bar{z} \vee x) \wedge (\bar{x} \vee y \vee z)$

Conjunction:  $x \leftrightarrow (y \wedge z) \Leftrightarrow (x \rightarrow y) \wedge (x \rightarrow z) \wedge ((y \wedge z) \rightarrow x)$   
 $\Leftrightarrow (\bar{x} \vee y) \wedge (\bar{x} \vee z) \wedge \overline{((y \wedge z) \vee x)}$   
 $\Leftrightarrow (\bar{x} \vee y) \wedge (\bar{x} \vee z) \wedge (\bar{y} \vee \bar{z} \vee x)$

Equivalence:  $x \leftrightarrow (y \leftrightarrow z) \Leftrightarrow (x \rightarrow (y \leftrightarrow z)) \wedge ((y \leftrightarrow z) \rightarrow x)$   
 $\Leftrightarrow (x \rightarrow ((y \rightarrow z) \wedge (z \rightarrow y))) \wedge ((y \leftrightarrow z) \rightarrow x)$   
 $\Leftrightarrow (x \rightarrow (y \rightarrow z)) \wedge (x \rightarrow (z \rightarrow y)) \wedge ((y \leftrightarrow z) \rightarrow x)$   
 $\Leftrightarrow (\bar{x} \vee \bar{y} \vee z) \wedge (\bar{x} \vee \bar{z} \vee y) \wedge ((y \leftrightarrow z) \rightarrow x)$   
 $\Leftrightarrow (\bar{x} \vee \bar{y} \vee z) \wedge (\bar{x} \vee \bar{z} \vee y) \wedge (((y \wedge z) \vee (\bar{y} \wedge \bar{z})) \rightarrow x)$   
 $\Leftrightarrow (\bar{x} \vee \bar{y} \vee z) \wedge (\bar{x} \vee \bar{z} \vee y) \wedge ((y \wedge z) \rightarrow x) \wedge ((\bar{y} \wedge \bar{z}) \rightarrow x)$   
 $\Leftrightarrow (\bar{x} \vee \bar{y} \vee z) \wedge (\bar{x} \vee \bar{z} \vee y) \wedge (\bar{y} \vee \bar{z} \vee x) \wedge (y \vee z \vee x)$

# Improving on Tseitin Encoding

- “cone of influence reduction” removes unrelated sub-formulas
- flatten associative operators (AND, OR) into multi-arity operators
  - without destroying sharing thus only applied to trees  $\Rightarrow$  variable elimination
- structural hashing of common sub-expression  $\Rightarrow$  AIGs
- switch to NNF based encoding close to leafs of the formula / circuit
  - as proposed by [Boy de la Tour'90]  $\Rightarrow$  variable elimination
  - but see also NiceDAGs [ChambersManoliosVroon'09]
- polarity based encoding
  - first described by [PlaistedGreenbaum'86]
  - can save half of the clauses  $\Rightarrow$  blocked clause elimination
- technolog based encoding [EénMishchenkoSörensson'07]
  - use heavy-weight circuit optimization for circuit chunks

**Most of these Optimizations can be achieved by Preprocessing!**

# Example for Checking Aliasing

original

```
assert (i != k);  
a[i] = a[k];  
a[j] = a[k];
```

optimized

```
int t = a[k];  
a[i] = t;  
a[j] = t;
```

$i \neq k$

$b_1 = \text{write}(a, i, t)$

$b_2 = \text{write}(b_1, j, s)$

$s = \text{read}(b_1, k)$

$t = \text{read}(a, k)$

$c_1 = \text{write}(a, i, t)$

$c_2 = \text{write}(c_1, j, t)$

static-single assignment (SSA)

original  $\neq$  optimized

iff

$b_2 \neq c_2$

$b_2 \neq c_2$

iff

$\exists l$

with

$\text{read}(b_2, l) \neq \text{read}(c_2, l)$

## Aliasing Example Continued 1

thus original  $\neq$  optimized iff

$i \neq k$

$t = \text{read}(a, k)$

$b_1 = \text{write}(a, i, t)$

$b_2 = \text{write}(b_1, j, s)$

$c_1 = \text{write}(a, i, t)$

$c_2 = \text{write}(c_1, j, t)$

$s = \text{read}(b_1, k)$

$\text{read}(b_2, l) \neq \text{read}(c_2, l)$

satisfiable

## Aliasing Example Continued 2

thus original  $\neq$  optimized iff

$i \neq k$

$t = \text{read}(a, k)$

$b_1 = \text{write}(a, i, t)$

$b_2 = \text{write}(b_1, j, s)$

$c_1 = \text{write}(a, i, t)$

$c_2 = \text{write}(c_1, j, t)$

$s = \text{read}(b_1, k)$

$u = \text{read}(b_2, l)$

$v = \text{read}(c_2, l)$

$u \neq v$

satisfiable



## Aliasing Example Continued 3

after eliminating  $c_2$

$i \neq k$

$t = \text{read}(a, k)$

$b_1 = \text{write}(a, i, t)$

$b_2 = \text{write}(b_1, j, s)$

$c_1 = \text{write}(a, i, t)$

$c_2 = \text{write}(c_1, j, t)$

$s = \text{read}(b_1, k)$

$u = \text{read}(b_2, l)$

$v = (i = j ? t : \text{read}(c_1, l))$

$u \neq v$

## Aliasing Example Continued 4

after eliminating  $c_2, c_1$

$i \neq k$

$t = \text{read}(a, k)$

$b_1 = \text{write}(a, i, t)$

$b_2 = \text{write}(b_1, j, s)$

$c_1 = \text{write}(a, i, t)$

$c_2 = \text{write}(c_1, j, t)$

$s = \text{read}(b_1, k)$

$u = \text{read}(b_2, l)$

$v = (l = j ? t : (l = i ? t : \text{read}(a, l)))$

$u \neq v$

## Aliasing Example Continued 5

after eliminating  $c_2, c_1, b_2$

$i \neq k$

$t = \text{read}(a, k)$

$b_1 = \text{write}(a, i, t)$

$b_2 = \text{write}(b_1, j, s)$

$c_1 = \text{write}(a, i, t)$

$c_2 = \text{write}(c_1, j, t)$

$s = \text{read}(b_1, k)$

$u = (l = j ? s : \text{read}(b_1, l))$

$v = (l = j ? t : (l = i ? t : \text{read}(a, l)))$

$u \neq v$

## Aliasing Example Continued 6

after eliminating  $c_2, c_1, b_2, b_1$

$i \neq k$

$t = \text{read}(a, k)$

$b_1 = \text{write}(a, i, t)$

$b_2 = \text{write}(b_1, j, s)$

$c_1 = \text{write}(a, i, t)$

$c_2 = \text{write}(c_1, j, t)$

$s = (k = i ? t : \text{read}(a, k))$

$u = (l = j ? s : (l = i ? t : \text{read}(a, l)))$

$v = (l = j ? t : (l = i ? t : \text{read}(a, l)))$

$u \neq v$

## Aliasing Example Continued 7

result after “write” elimination

$i \neq k$

$t = \text{read}(a, k)$

$s = (k = i ? t : \text{read}(a, k))$

$u = (l = j ? s : (l = i ? t : \text{read}(a, l)))$

$v = (l = j ? t : (l = i ? t : \text{read}(a, l)))$

$u \neq v$

## Aliasing Example Continued 8

after eliminating conditionals (if-then-else)

$$i \neq k$$
$$t = \text{read}(a, k)$$
$$k = i \rightarrow s = t$$
$$k \neq i \rightarrow s = \text{read}(a, k)$$
$$l = j \rightarrow u = s$$
$$l \neq j \wedge l = i \rightarrow u = t$$
$$l \neq j \wedge l \neq i \rightarrow u = \text{read}(a, l)$$
$$l = j \rightarrow v = t$$
$$l \neq j \wedge l = i \rightarrow v = t$$
$$l \neq j \wedge l \neq i \rightarrow v = \text{read}(a, l)$$
$$u \neq v$$

now treat “read” as uninterpreted function (say  $f$ )

## Aliasing Example Continued 9

after “Ackermanization” using  $x = \text{read}(a, k)$ ,  $y = \text{read}(a, l)$

$$i \neq k$$

$$t = x$$

$$k = i \rightarrow s = t$$

$$k \neq i \rightarrow s = x$$

$$l = j \rightarrow u = s$$

$$l \neq j \wedge l = i \rightarrow u = t$$

$$l \neq j \wedge l \neq i \rightarrow u = y$$

$$l = j \rightarrow v = t$$

$$l \neq j \wedge l = i \rightarrow v = t$$

$$l \neq j \wedge l \neq i \rightarrow v = y$$

$$u \neq v$$

$$k = l \rightarrow x = y$$

10 variables remain

use 4-bit bitvectors to encode 0..15

# Bit-Blasting of Bit-Vector Equality

equality  $x = y$  of 4-bit bitvectors  $x, y$  as new literal  $\ell_{x=y}$

$$[x_3, x_2, x_1, x_0]_4 = [y_3, y_2, y_1, y_0]_4$$

$$\ell_{x=y} \iff \bigwedge_{i=0}^3 x_i \iff y_i$$

and then use Tseitin encoding



# Bit-Blasting of Bit-Vector Addition

addition of 4-bit numbers  $x, y$  with result  $s$  also 4-bit:  $s = x + y$

$$[s_3, s_2, s_1, s_0]_4 = [x_3, x_2, x_1, x_0]_4 + [y_3, y_2, y_1, y_0]_4$$

$$[s_3, \cdot]_2 = \text{FullAdder}(x_3, y_3, c_2)$$

$$[s_2, c_2]_2 = \text{FullAdder}(x_2, y_2, c_1)$$

$$[s_1, c_1]_2 = \text{FullAdder}(x_1, y_1, c_0)$$

$$[s_0, c_0]_2 = \text{FullAdder}(x_0, y_0, \text{false})$$

where

$$[s, o]_2 = \text{FullAdder}(x, y, i) \quad \text{with}$$

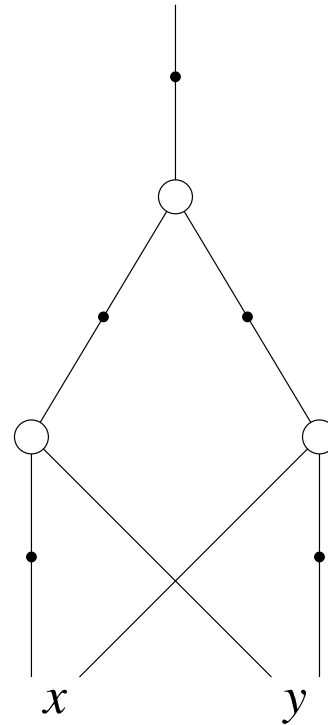
$$s = x \text{ xor } y \text{ xor } i$$

$$o = (x \wedge y) \vee (x \wedge i) \vee (y \wedge i) = ((x + y + i) \geq 2)$$

# Intermediate Representations

- encoding directly into CNF is hard, so we use intermediate levels:
  1. application level (SSA, encoding execution semantics)
  2. bit-precise semantics world-level operations (bit-vectors)
  3. bit-level representations such as And-Inverter Graphs (AIGs)
  4. conjunctive normal form (CNF)
- encoding “logical” constraints is another story

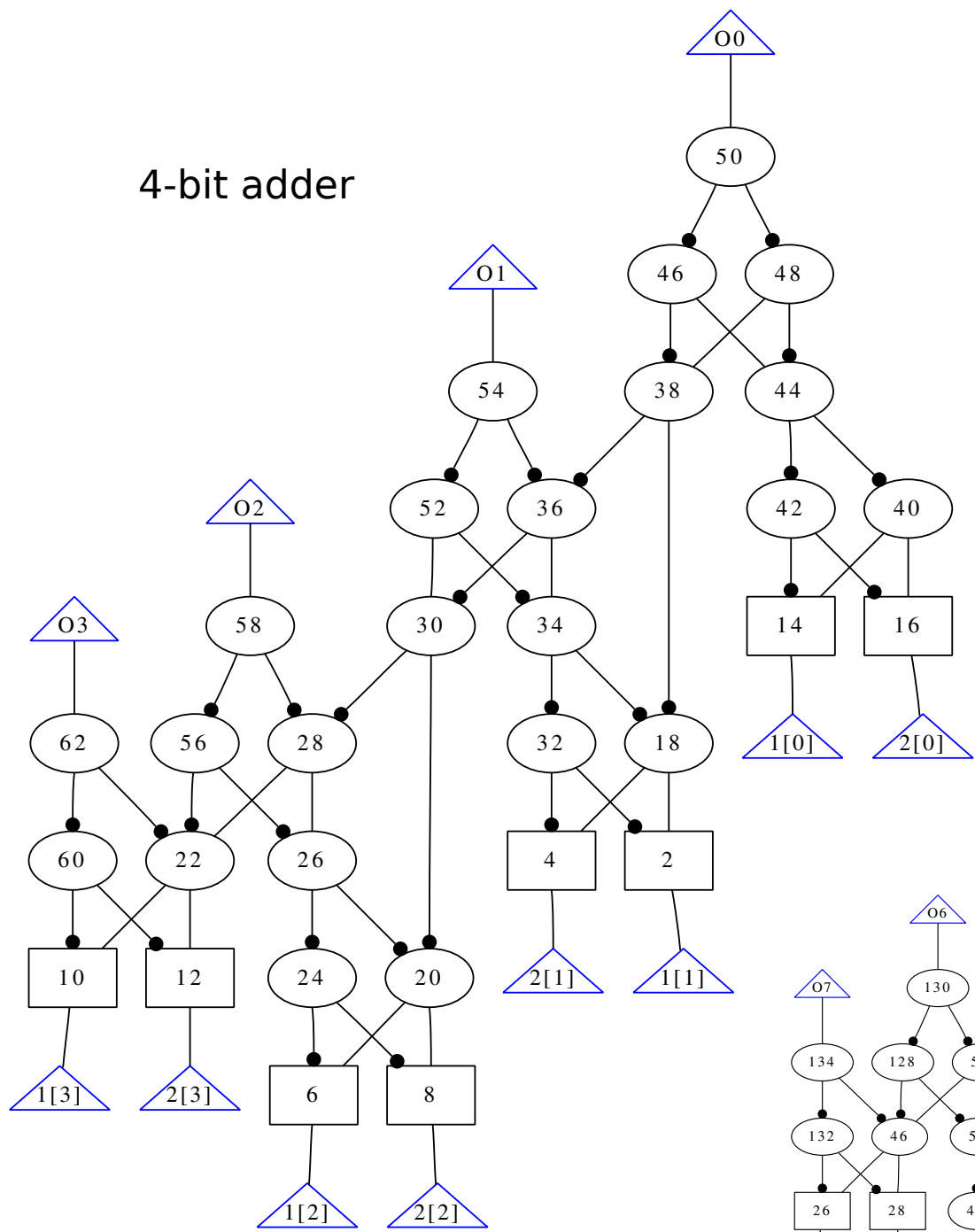
# XOR as AIG



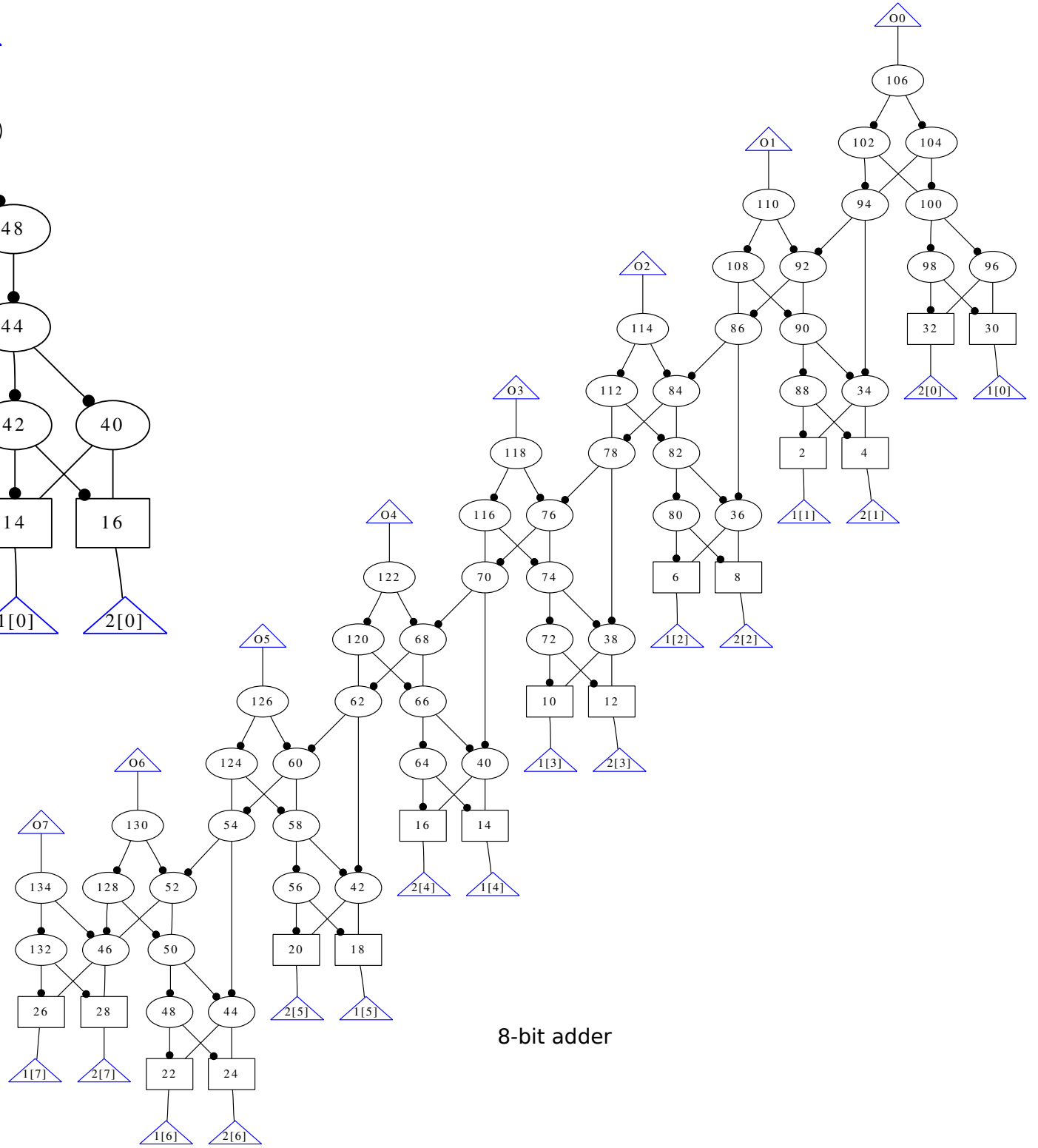
negation/sign are edge attributes  
not part of node

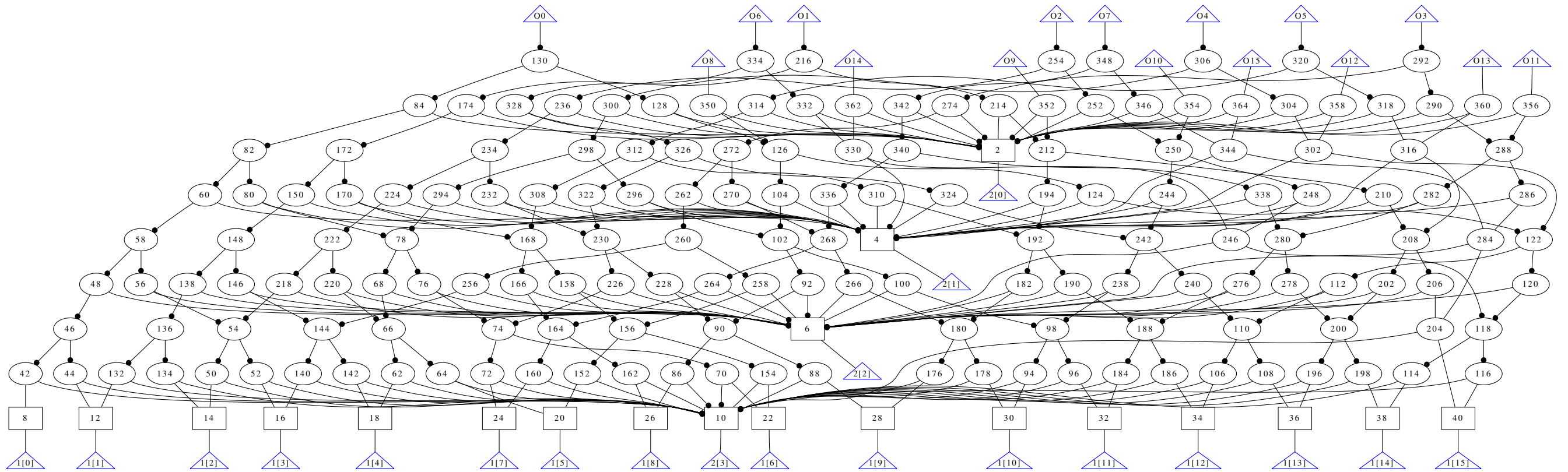
$$x \text{ xor } y \equiv (\bar{x} \wedge y) \vee (x \wedge \bar{y}) \equiv \overline{\overline{(\bar{x} \wedge y)} \wedge \overline{(x \wedge \bar{y})}}$$

4-bit adder

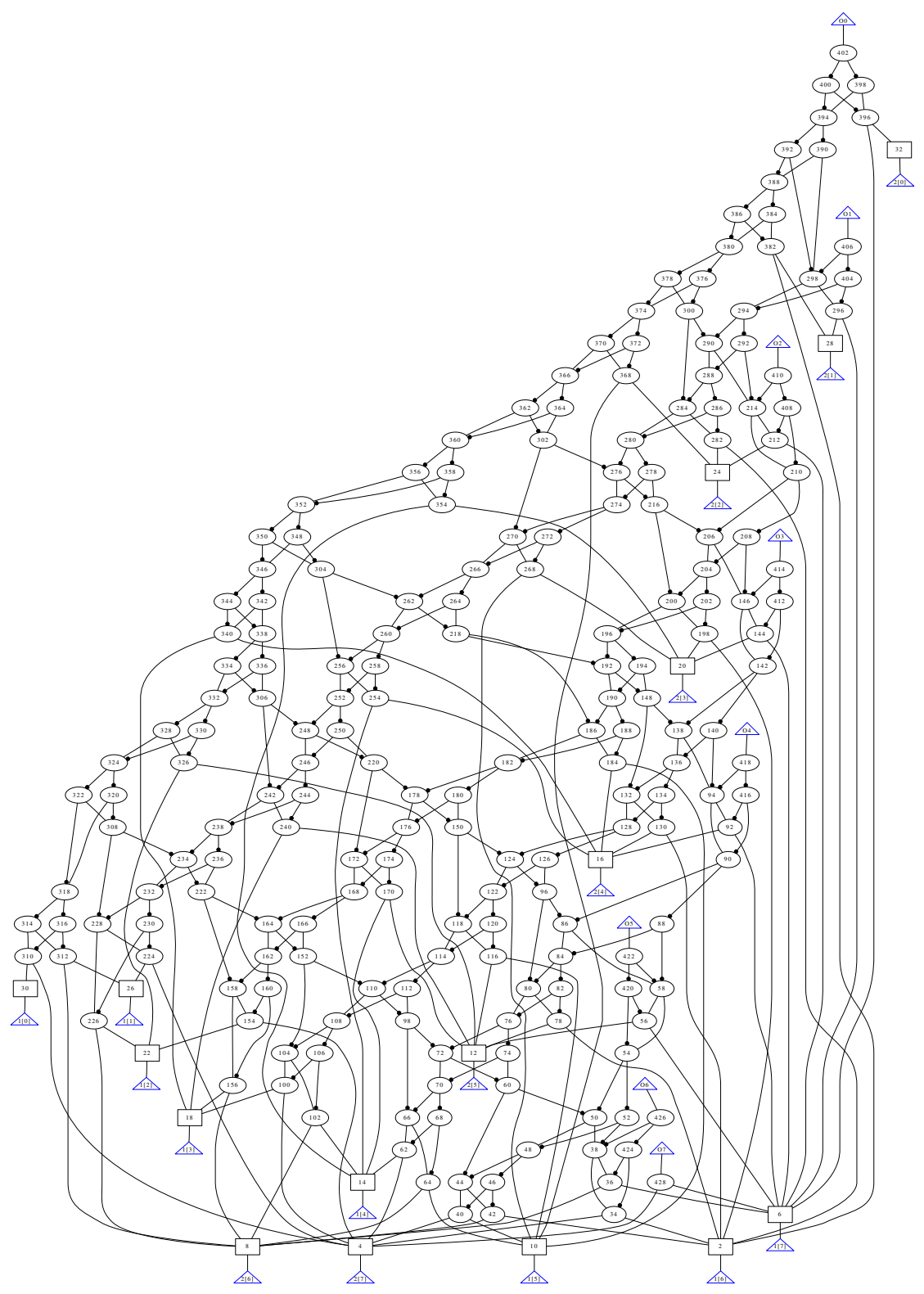


8-bit adder





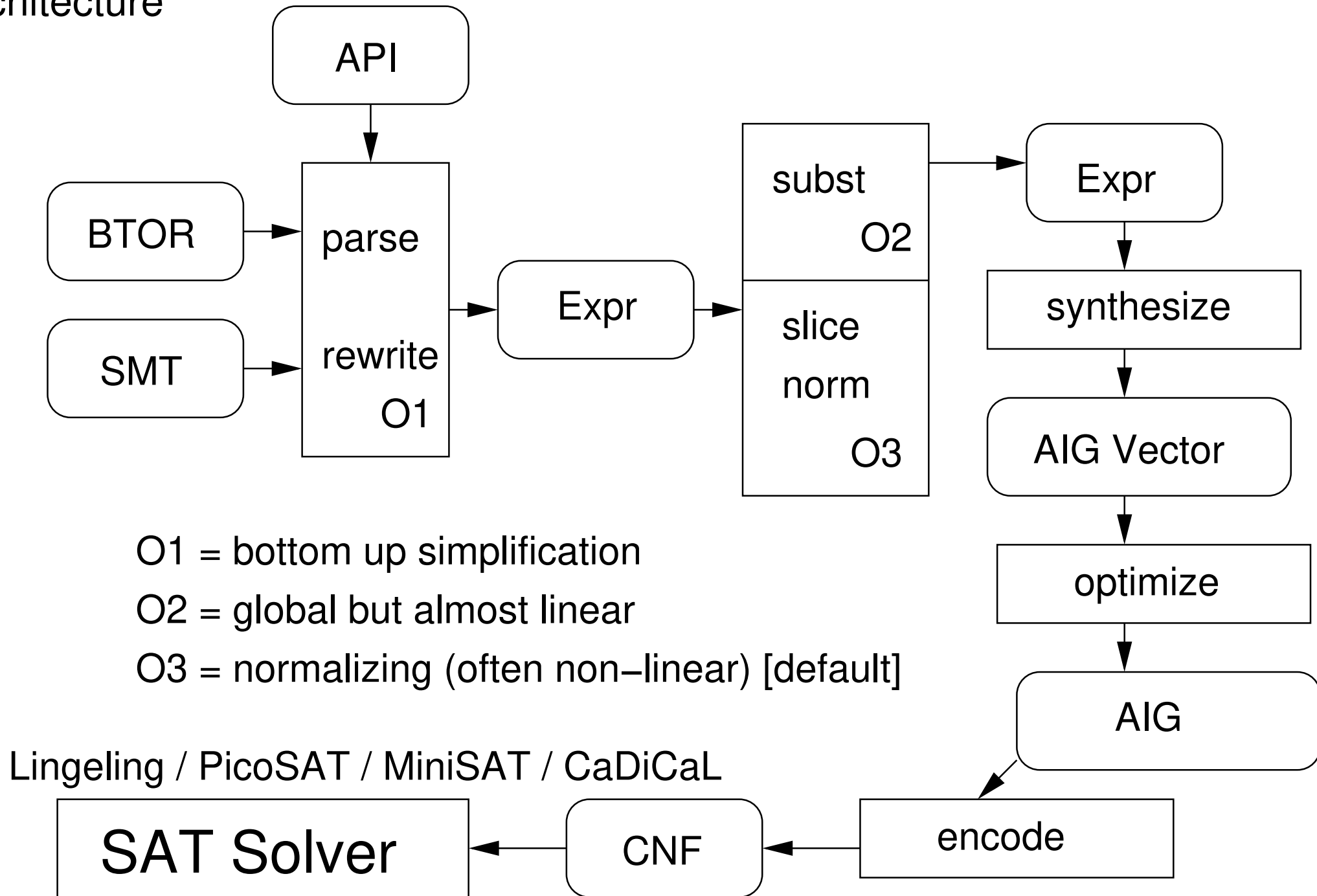
bit-vector of length 16 shifted by bit-vector of length 4



# Complexity of Bit-Blasting

- can handle arbitrary strange computations
  - in essence anything your computer can compute
  - like non-linear constraints, division, modulo, bit-wise operators, shift, ...
  - floating points (well ... see Martin Brain's talk tomorrow)
  - can make full use of power of SAT solvers (including preprocessing)
- sometimes generates big and hard SAT formulas
  - deciding bit-vector logic NEXPTIME complete [[KovácsnaiFröhlichBiere'12](#)]
  - since bit-width is encoded logarithmically
  - even one 32-bit multiplication needs 2000+ AIG nodes
- software engineering for bit-blasting is tricky
  - SAT solvers not good at structural hashing [[HeuleJärvisaloBiere'13](#)] [[HeuleBiere'13](#)]
  - so need to maintain an intermediate format like AIGs
  - not easy in an incremental setting (inprocessing?)

# Boolector Architecture

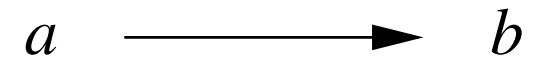




# Encoding Logical Constraints

- Tseitin construction suitable for most kinds of “model constraints”
  - assuming simple operational semantics: encode an interpreter
  - small domains: one-hot encoding      large domains: binary encoding  
check out “order encoding” too
- harder to encode properties or additional constraints
  - temporal logic / fix-points
  - environment constraints
- example for fix-points / recursive equations:  $x = (a \vee y), \quad y = (b \vee x)$ 
  - has unique least fix-point  $x = y = (a \vee b)$
  - and unique largest fix-point  $x = y = true$  but unfortunately ...
  - ... only largest fix-point can be (directly) encoded in SAT
  - otherwise need stable models / logical programming / ASP

# Encoding Reachability in Prolog for Graph with 2 Nodes



```
edge(a,b) .
```

```
reach(X,Y) :- edge(X,Y) .
```

```
reach(X,Y) :- edge(X,Z) , reach(Z,Y) .
```

```
?- reach(b,a) .
```

# Wrong SAT Encoding for Graph with 2 Nodes

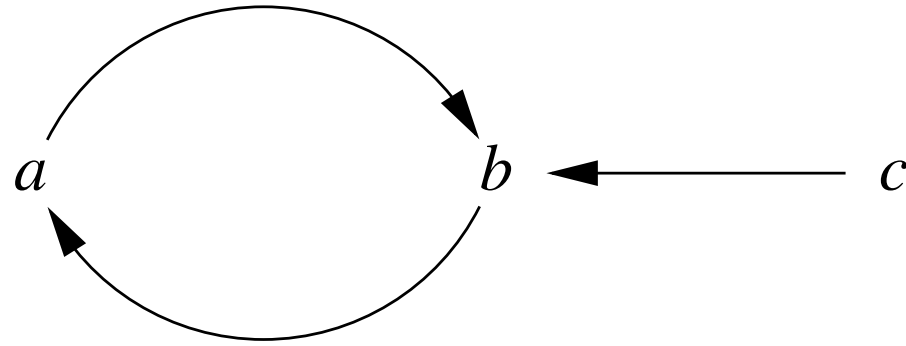
```
edge_a_b &
(reach_a_a <- edge_a_a) &
(reach_a_b <- edge_a_b) &
(reach_b_a <- edge_b_a) &
(reach_b_b <- edge_b_b) &
(reach_a_a <- edge_a_a & reach_a_a) &
(reach_a_a <- edge_a_b & reach_b_a) &
(reach_a_b <- edge_a_a & reach_a_b) &
(reach_a_b <- edge_a_b & reach_b_b) &
(reach_b_a <- edge_b_a & reach_a_a) &
(reach_b_a <- edge_b_b & reach_b_a) &
(reach_b_b <- edge_b_a & reach_a_b) &
(reach_b_b <- edge_b_b & reach_b_b) &
reach_b_a
```

# Right SAT Encoding for Graph with 2 Nodes

```
!edge_a_a &      (reach_a_a_1 <-> (edge_a_a | (edge_a_a & reach_a_a_0))) &
edge_a_b &      (reach_a_a_1 <-> (edge_a_a | (edge_a_b & reach_b_a_0))) &
!edge_b_a &      (reach_a_b_1 <-> (edge_a_b | (edge_a_a & reach_a_b_0))) &
!edge_b_b &      (reach_a_b_1 <-> (edge_a_b | (edge_a_b & reach_b_b_0))) &
                 (reach_b_a_1 <-> (edge_b_a | (edge_b_a & reach_a_a_0))) &
!reach_a_a_0 &  (reach_b_a_1 <-> (edge_b_a | (edge_b_b & reach_b_a_0))) &
!reach_a_b_0 &  (reach_b_b_1 <-> (edge_b_b | (edge_b_a & reach_a_b_0))) &
!reach_b_a_0 &  (reach_b_b_1 <-> (edge_b_b | (edge_b_b & reach_b_b_0))) &
!reach_b_b_0 &  (reach_a_a_2 <-> (edge_a_a | (edge_a_a & reach_a_a_1))) &
                 (reach_a_a_2 <-> (edge_a_a | (edge_a_b & reach_b_a_1))) &
                 (reach_a_b_2 <-> (edge_a_b | (edge_a_a & reach_a_b_1))) &
                 (reach_a_b_2 <-> (edge_a_b | (edge_a_b & reach_b_b_1))) &
                 (reach_b_a_2 <-> (edge_b_a | (edge_b_a & reach_a_a_1))) &
                 (reach_b_a_2 <-> (edge_b_a | (edge_b_b & reach_b_a_1))) &
                 (reach_b_b_2 <-> (edge_b_b | (edge_b_a & reach_a_b_1))) &
                 (reach_b_b_2 <-> (edge_b_b | (edge_b_b & reach_b_b_1))) &

(reach_b_a_0 | reach_b_a_1 | reach_b_a_2)
```

# Encoding Reachability in Prolog for Graph with 3 Nodes



edge (a, b) .

edge (b, a) .

edge (c, b) .

reach (X, Y) :- edge (X, Y) .

reach (X, Y) :- edge (X, Z) , reach (Z, Y) .

?- reach (a, c) .

# Wrong SAT Encoding for Graph with 3 Nodes

```
edge_a_b & (reach_a_a <- edge_a_a & reach_a_a) &
edge_b_a & (reach_a_a <- edge_a_b & reach_b_a) &
edge_c_b & (reach_a_a <- edge_a_c & reach_c_a) &
           (reach_a_b <- edge_a_a & reach_a_b) &
(reach_a_a <- edge_a_a) & (reach_a_b <- edge_a_b & reach_b_b) & (reach_c_a <- edge_c_a & reach_a_a) &
(reach_a_b <- edge_a_b) & (reach_a_b <- edge_a_c & reach_c_b) & (reach_c_a <- edge_c_b & reach_b_a) &
(reach_a_c <- edge_a_c) & (reach_a_c <- edge_a_a & reach_a_c) & (reach_c_a <- edge_c_c & reach_c_a) &
(reach_b_a <- edge_b_a) & (reach_a_c <- edge_a_b & reach_b_c) & (reach_c_b <- edge_c_a & reach_a_b) &
(reach_b_b <- edge_b_b) & (reach_a_c <- edge_a_c & reach_c_c) & (reach_c_b <- edge_c_b & reach_b_b) &
(reach_b_c <- edge_b_c) & (reach_b_a <- edge_b_a & reach_a_a) & (reach_c_b <- edge_c_c & reach_c_b) &
(reach_c_a <- edge_c_a) & (reach_b_a <- edge_b_b & reach_b_a) & (reach_c_c <- edge_c_a & reach_a_c) &
(reach_c_b <- edge_c_b) & (reach_b_a <- edge_b_c & reach_c_a) & (reach_c_c <- edge_c_b & reach_b_c) &
(reach_c_c <- edge_c_c) & (reach_b_b <- edge_b_a & reach_a_b) & (reach_c_c <- edge_c_c & reach_c_c) &
           (reach_b_b <- edge_b_b & reach_b_b) &
           (reach_b_b <- edge_b_c & reach_c_b) &
           (reach_b_c <- edge_b_a & reach_a_c) &
           (reach_b_c <- edge_b_b & reach_b_c) &
           (reach_b_c <- edge_b_c & reach_c_c) &
reach_a_c
```



# Encoding Least Fix-Points

- incremental encoding for least fix-points [[GebserKaufmannNeumannSchaub'07](#)]
  - use the “wrong encoding” and call SAT solver
  - if unsatisfiable no least fix-point exists
  - if satisfiable check solution for cyclic dependencies
  - if there is no cyclic dependency then the model is a least fix-point
  - otherwise add clause which removes cycle and continue
- other incremental encodings
  - simple path constraints in BMC /  $k$ -induction [[EénSörensson'03](#)]
  - lazy clause encoding in CP [[OhrimenkoStuckeyCodish'07](#)]
  - lemmas on demand for SMT [[deMouraRueß'02](#)] [[BrummayerBiere'09](#)]
- lazy encodings might result in adding exponential many clauses
- encoding temporal properties (LTL) for BMC [[LatvalaBiereHeljankoJuntilla'04](#)]
  - temporal operators with least fix-point semantics:  $Fp$ ,  $p \cup q$
  - needs only two “iterations” due to monotonicity of the semantics



## Example of Logical Constraints: Cardinality Constraints

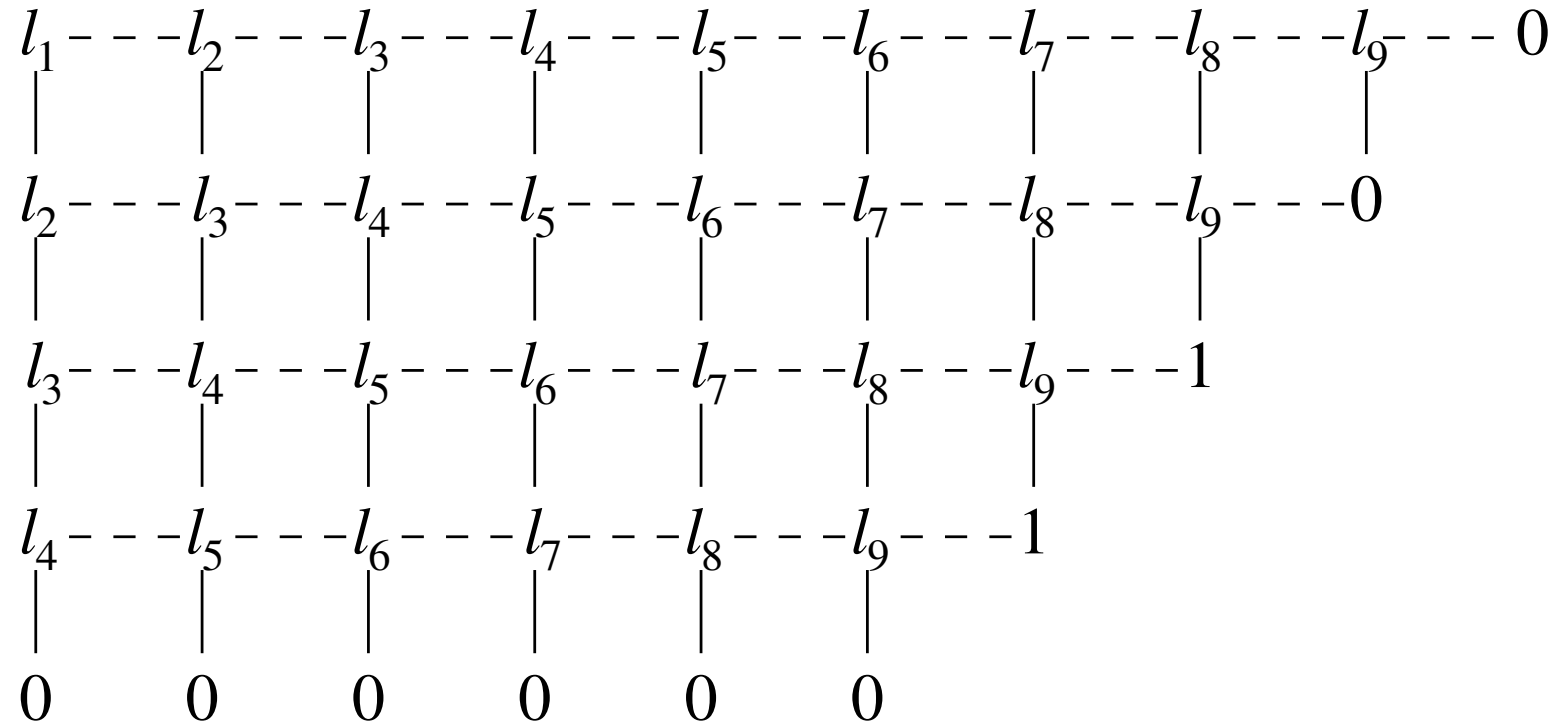
- given a set of literals  $\{l_1, \dots, l_n\}$ 
  - constraint the number of literals assigned to *true*
  - $l_1 + \dots + l_n \geq k$  or  $l_1 + \dots + l_n \leq k$  or  $l_1 + \dots + l_n = k$
  - combined make up exactly all fully symmetric boolean functions
- multiple encodings of cardinality constraints
  - naïve encoding exponential: at-most-one quadratic, at-most-two cubic, etc.
  - quadratic  $O(k \cdot n)$  encoding has its roots in [\[Shannon'38\]](#)
  - linear  $O(n)$  parallel counter encoding [\[Sinz'05\]](#)
- many variants even for at-most-one constraints
  - see [\[BiereLeBerreLoncaManthey'14\]](#) [\[MantheyHeuleBiere'13\]](#) for references
- Pseudo-Boolean constraints (PB) or 0/1 ILP constraints have many encodings too

$$2 \cdot \bar{a} + \bar{b} + c + \bar{d} + 2 \cdot e \geq 3$$

actually used to handle MaxSAT in SAT4J for configuration in Eclipse

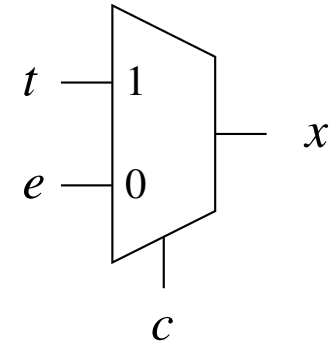
# BDD-Based Encoding of Cardinality Constraints

$$2 \leq l_1 + \dots + l_9 \leq 3$$



If-Then-Else gates (MUX) with “then” edge downward, dashed “else” edge to the right

# Tseitin Encoding of If-Then-Else Gate and Arc Consistency



$$\begin{aligned}x \leftrightarrow (c ? t : e) &\Leftrightarrow (x \rightarrow (c \rightarrow t)) \wedge (x \rightarrow (\bar{c} \rightarrow e)) \wedge (\bar{x} \rightarrow (c \rightarrow \bar{t})) \wedge (\bar{x} \rightarrow (\bar{c} \rightarrow \bar{e})) \\ &\Leftrightarrow (\bar{x} \vee \bar{c} \vee t) \wedge (\bar{x} \vee c \vee e) \wedge (x \vee \bar{c} \vee \bar{t}) \wedge (x \vee c \vee \bar{e})\end{aligned}$$

this is a minimal size CNF but the CNF is not **arc consistent**

- if  $t$  and  $e$  have the same value then  $x$  needs to have that too

$$(\bar{t} \wedge \bar{e} \rightarrow \bar{x}) \equiv (t \vee e \vee \bar{x}) \qquad (t \wedge e \rightarrow x) \equiv (\bar{t} \vee \bar{e} \vee x)$$

- but can be learned or derived through preprocessing (ternary resolution)  
keeping those clauses redundant is better in practice

# DIMACS Format

```
$ cat example.cnf
```

```
c comments start with 'c' and extend until the end of the line
```

```
c
```

```
c variables are encoded as integers:
```

```
c
```

```
c 'tie' becomes '1'
```

```
c 'shirt' becomes '2'
```

```
c
```

```
c header 'p cnf <variables> <clauses>'
```

```
c
```

```
p cnf 2 3
```

```
-1 2 0          c !tie or shirt
```

```
1 2 0          c tie or shirt
```

```
-1 -2 0        c !tie or !shirt
```

```
$ picosat example.cnf
```

```
s SATISFIABLE
```

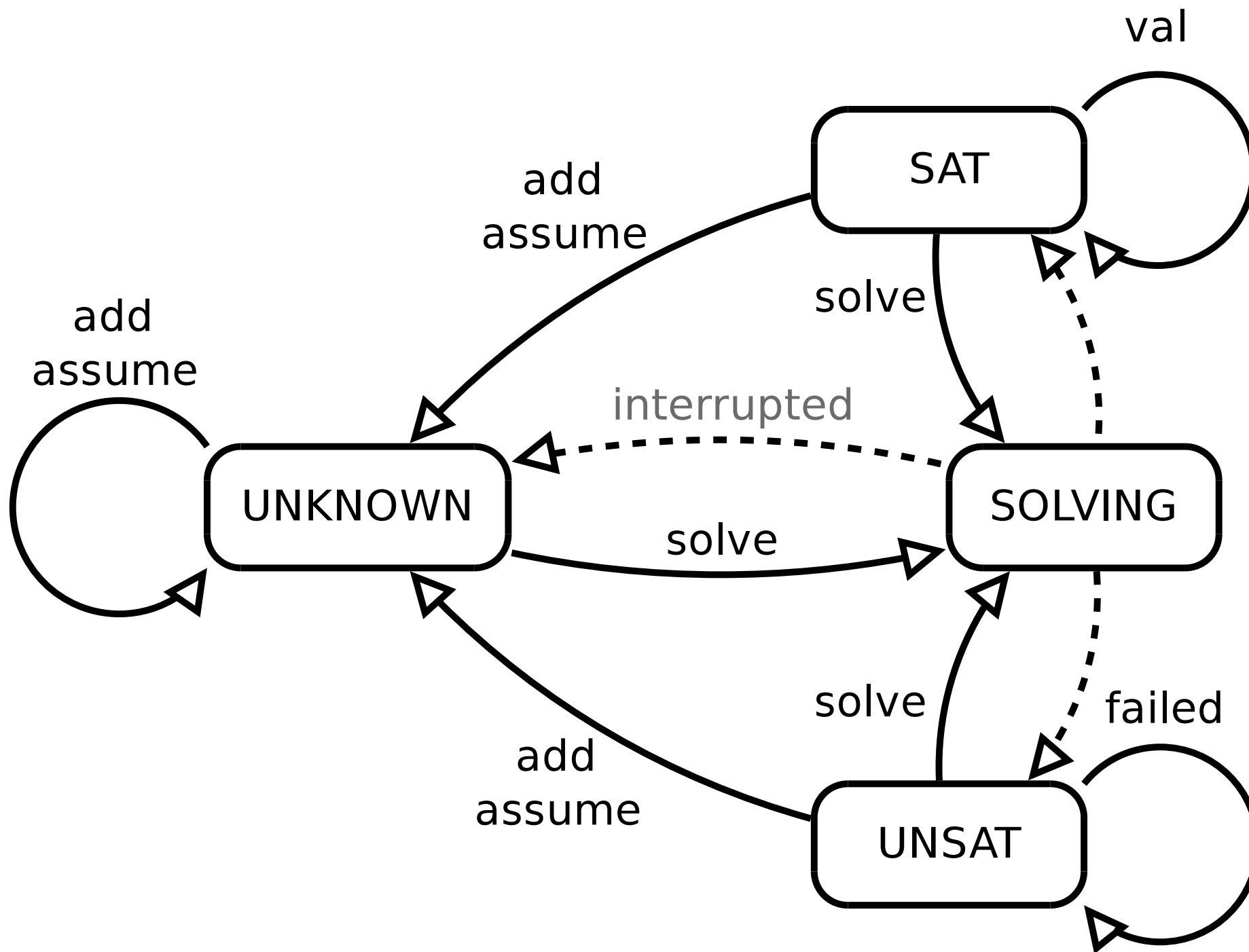
```
v -1 2 0
```

# SAT Application Programmatic Interface (API)

- incremental usage of SAT solvers
  - add facts such as clauses incrementally
  - call SAT solver and get satisfying assignments
  - optionally retract facts UNSAT, UNSAT, ... , UNSAT, SAT vs. SAT, SAT, ... , UNSAT
- retracting facts
  - remove clauses explicitly: complex to implement
  - push / pop: stack like activation, no sharing of learned facts (as in SMTLIB)
  - MiniSAT assumptions [\[EénSörensson'03\]](#)
- assumptions
  - unit assumptions: assumed for the next SAT call
  - easy to implement: force SAT solver to decide on assumptions first
  - shares learned clauses across SAT calls
- IPASIR: Reentrant Incremental SAT API
  - used in the SAT competition / race since 2015

[\[BalyoBierelserSinz'16\]](#)

# IPASIR Model



```

#include "ipasir.h"
#include <assert.h>
#include <stdio.h>
#define ADD(LIT) ipasir_add (solver, LIT)
#define PRINT(LIT) \
    printf (ipasir_val (solver, LIT) < 0 ? " -" #LIT : " " #LIT)
int main () {
    void * solver = ipasir_init ();
    enum { tie = 1, shirt = 2 };
    ADD (-tie); ADD ( shirt); ADD (0);
    ADD ( tie); ADD ( shirt); ADD (0);
    ADD (-tie); ADD (-shirt); ADD (0);
    int res = ipasir_solve (solver);
    assert (res == 10);
    printf ("satisfiable:"); PRINT (shirt); PRINT (tie); printf ("\n");
    printf ("assuming now: tie shirt\n");
    ipasir_assume (solver, tie); ipasir_assume (solver, shirt);
    res = ipasir_solve (solver);
    assert (res == 20);
    printf ("unsatisfiable, failed:");
    if (ipasir_failed (solver, tie)) printf (" tie");
    if (ipasir_failed (solver, shirt)) printf (" shirt");
    printf ("\n");
    ipasir_release (solver);
    return res;
}

```

```

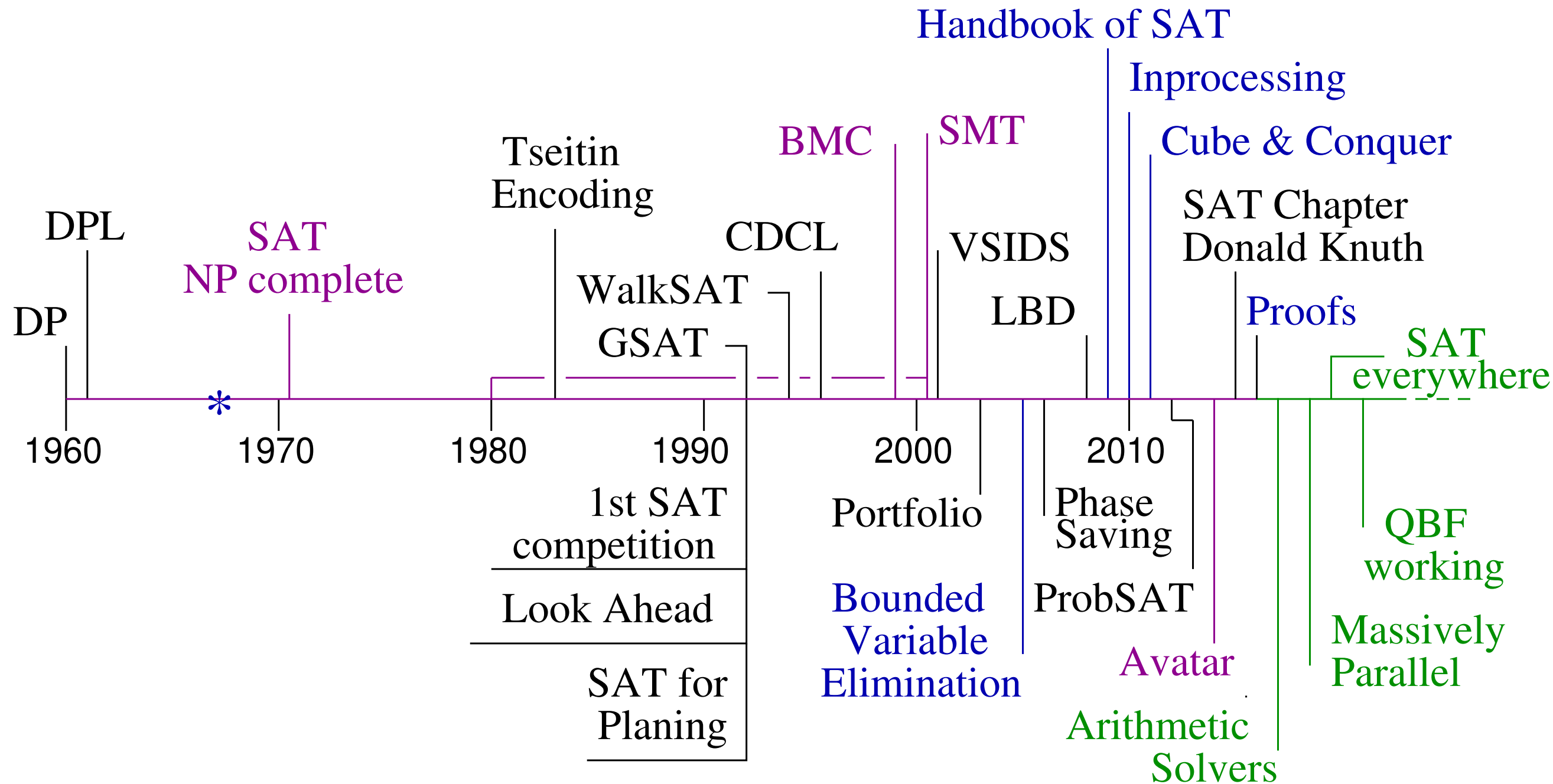
$ ./example
satisfiable: shirt -tie
assuming now: tie shirt
unsatisfiable, failed: tie

```





# Personal SAT Solver History



## Links

- <https://fmv.jku.at/limboole>
- <https://fmv.jku.at/aiger>
- <https://github.com/Boolector/boolector>
- <https://github.com/biotomas/ipasir>
- <https://github.com/arminbiere/cadical>
- <https://github.com/arminbiere/lingeling>

## Jobs

- new LIT AI Lab
  - Linz Institute of Technology (LIT)
  - Artificial Intelligence (AI)
- new LIT AI PhD School
- world-class experts
  - machine learning Hochreiter, Widmer
  - SAT / SMT / AR Biere, Seidl, Kauers
- deductive & inductive reasoning

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