

Design of an environment for solving pseudo-Boolean optimization problems

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Introduction

- C++ library
- Reduce time to solve pseudo-Boolean minimisation problems
- Build onto an existing one, the *Base Project*
- Pseudo-Boolean constraints are encoded with *PBLib*
- Two search algorithms to find the optimal value
- Two timeout strategies

Basic concepts

Boolean formula

- Variables

$$a, b, c, \dots \in \mathbb{B}$$

- Connectors

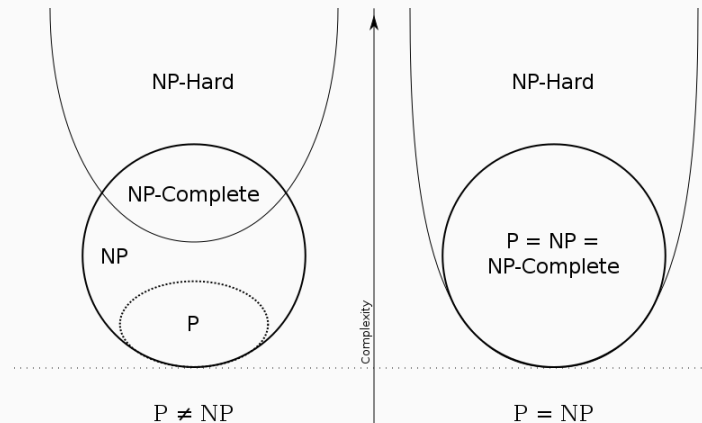
$$\wedge, \vee, \neg$$

- For exemple

$$a \wedge b \vee (\neg c)$$

Boolean satisfiability problem

- Given a propositional formula f , is there a truth assignment i such that $i(f) = 1$?
- NP-complete



Conjunctive Normal Form

- Clause: Finite disjunction of literals

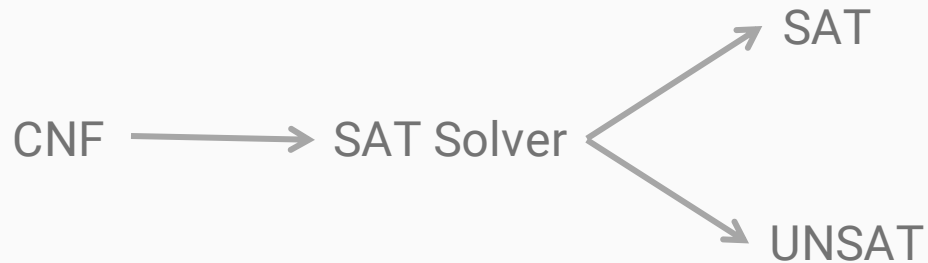
$$l_1 \vee \dots \vee l_n$$

- CNF: Conjunction of one or more clauses

$$(A \vee B) \wedge C$$

SAT Solver

- Software for solving SAT



Example

- Boolean formula:

$$a \vee (b \wedge c)$$

- CNF:

$$(a \vee b) \wedge (a \vee c)$$

- Solver

Satisfiable

$$I_1(a) = 1, I_1(b) = 0, I_1(c) = 0$$

Base project

Motivation

- Existing encoding to CNF
- New transformation

- Reduce the number of clauses adding auxiliary variables
- See the tradeoffs

Requirements

- Straightforward notation: literals and operators

```
1 Formula a = BoolFunc::newLit("a");  
2 Formula b = BoolFunc::newLit("b");  
3  
4 Formula f = (a+b) * b;
```

Transformations

```
Formula f = (a+b)*b;  
Cnf cnf = CnfConverter::tseytin(f);
```

- Tseytin

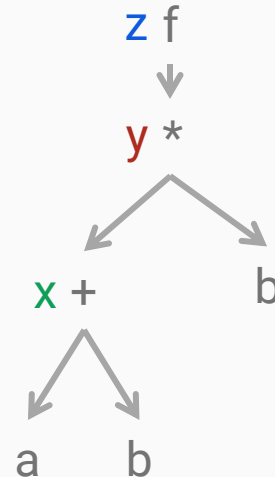
- Length CNF is linear in the size of the formula
- Equisatisfiable

$$x = a \vee b$$

$$a \rightarrow x, b \rightarrow x, x \rightarrow (a \vee b)$$

$$y = x \wedge b$$

$$y \rightarrow x, y \rightarrow b, (x \wedge b) \rightarrow y$$



Transformations

```
Formula f = (a+b)*b;
//Convert formula to bdd
Cnf cnf = CnfConverter::convertToCnf(f_bdd);
```

- Extracting primes

$$f = (a * b) + (c * d)$$

$$cnf = (a + c) * (a + d) * (b + c) * (b + d)$$

$$\bar{f} = (\bar{a} + \bar{b}) * (\bar{c} + \bar{d})$$

$$SoP = (\bar{a} * \bar{c}) + (\bar{a} * \bar{d}) \\ + (\bar{b} * \bar{c}) + (\bar{b} * \bar{d})$$

- Cudd - Colorado University Decision Diagram package

pc \ ab	00	01	11	10
00	0	0	1	0
01	0	0	1	0
11	1	1	1	1
10	0	0	1	0

pc \ ab	00	01	11	10
00	1	1	0	1
01	1	1	0	1
11	0	0	0	0
10	1	1	0	1

Transformations

- Problem

$$f = \text{XOR}(a, b, c, d)$$

$$\begin{aligned} \text{cnf} = & \overline{abcd} + \overline{abcd} \\ & + \overline{ab}cd + \overline{abcd} \\ & + \overline{abcd} + \overline{abcd} \\ & + \overline{abcd} + abcd \end{aligned}$$

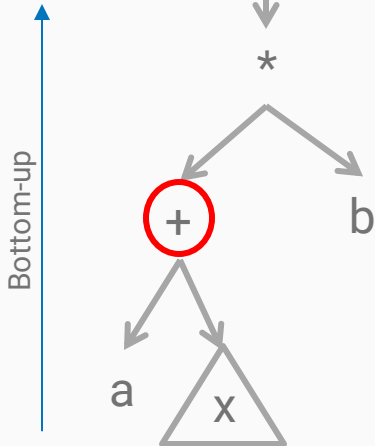
- Idea: When function's primes cover a small space then a lot of clauses are required

cd \ ab	00	01	11	10
00	1	0	1	0
01	0	1	0	1
11	1	0	1	0
10	0	1	0	1

Transformations

```
Formula f = (a+b)*b;  
MixCNFConverter m = MixCNFConverter();  
m.convert(f);  
Cnf cnf = m.getResult();
```

- Extracting primes and adding new variables

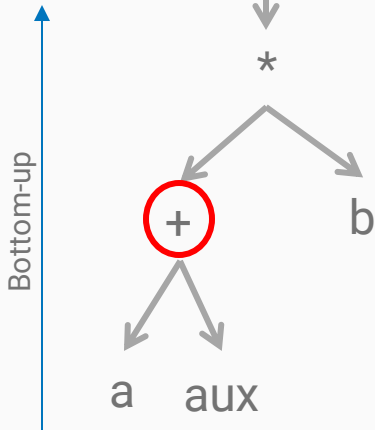


- Build BDD subtree
- Look at the largest cube ratio
- If not big enough
 - Get biggest child (x)
- If big enough generate CNF extracting primes

Transformations

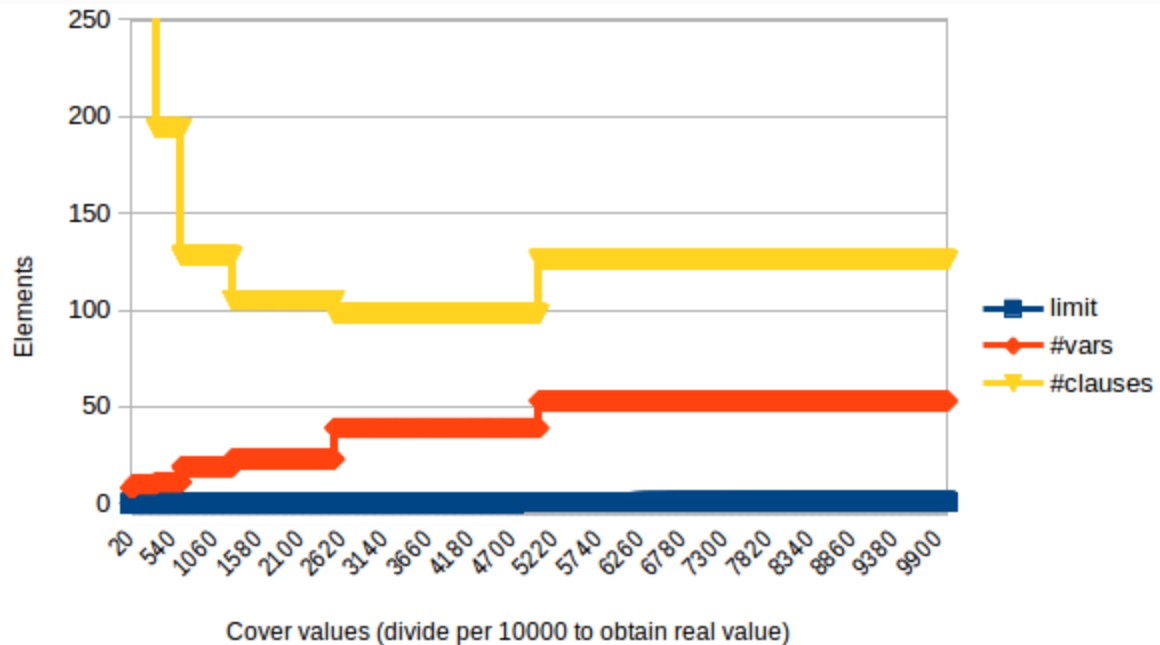
```
Formula f = (a+b)*b;  
MixCNFConverter m = MixCNFConverter();  
m.convert(f);  
Cnf cnf = m.getResult();
```

- Extracting primes and adding new variables



- Build BDD subtree
- Look at the largest cube ratio
- If not big enough
 - Get biggest child (x)
 - Replace it by an auxiliary variable
 - Add Δx XNOR aux
- If big enough generate CNF extracting primes
- Continue with the parent node

Experiment



The problem: Pseudo-Boolean optimisation

Pseudo-Boolean optimisation

- Pseudo-Boolean constraints

$$w_1x_1 + w_2x_2 + \dots + w_nx_n \# k$$

$$w_i, k \in \mathbb{Z}$$

$$x_i \in \mathbb{B}$$

$$\# \in \{=, \leq, \geq, <, >\}$$

- Cost function

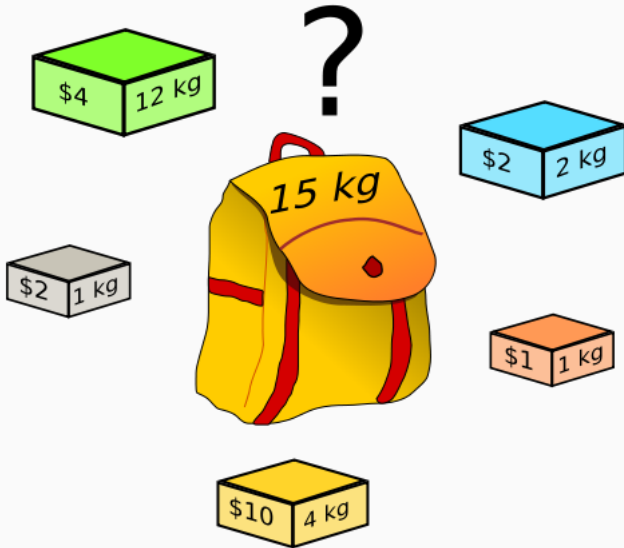
$$w_1x_1 + w_2x_2 + \dots + w_nx_n$$

$$w_i \in \mathbb{Z}$$

$$x_i \in \mathbb{B}$$

maximize	$\mathbf{c}^T \mathbf{x}$
subject to	$A\mathbf{x} \leq \mathbf{b}$
and	$\mathbf{x} \geq \mathbf{0}$

Example - Knapsack problem



Variables:

$$o_1, o_2, \dots, o_n \in \mathbb{B}$$

$$w_1, w_2, \dots, w_n \in \mathbb{Z}$$

$$v_1, v_2, \dots, v_n \in \mathbb{Z}$$

Constraints:

$$w_1 o_1 + w_2 o_2 + \dots + w_n o_n \leq$$

knapsack's capacity

Cost function:

$$v_1 o_1 + v_2 o_2 + \dots + v_n o_n$$

TFG software

Objectives

- Pseudo-Boolean minimisation
- Timeouts
- Multithreading*

PBLib

- C++ library
- Encodings maintain arc consistency by unit propagation
- Decides which encoder provides the most effective translation
- Variable's management done by the user

<i>At most one</i>	<i>At most K</i>	<i>PB</i>
sequential*	BDD**	BDD
bimander	cardinality networks	adder networks
commander	adder networks	watchdog
k-product	todo: perfect hashing	sorting networks
binary		binary merge
pairwise		sequential weight counter
nested		

* equivalent to BDD, latter and regular encoding
** equivalent to sequential counter
Encodings labeled with *todo* are planned for the (near) future.

Pseudo-Boolean minimisation layer

- Variables *int32_t*
 - Coefficients *int64_t*
 - Compact representation
- } Maximum compatibility with PBLib

```
PBFormula pf = PBFormula({-3,2}, {1,-2});
```

$$-3x_1 + 2\overline{x_2}$$

```
PBConstraint c = PBConstraint(PBFormula({3,2},{1,2}),1);
```

$$3x_1 + 2x_2 \leq 1$$

Pseudo-Boolean minimisation layer

```
std::vector< PBConstraint > constraints = {  
    PBConstraint(PBFormula({1,2},{-1,-2}),1),  
    PBConstraint(PBFormula({3,4},{2,-3}),1),  
    PBConstraint(PBFormula({3,7},{1,-3}),1)  
};  
  
PBMin m = PBMin(constraints, PBFormula({3,-5},{4,5}));
```

$$\overline{x_1} - 2\overline{x_2} \leq 1,$$

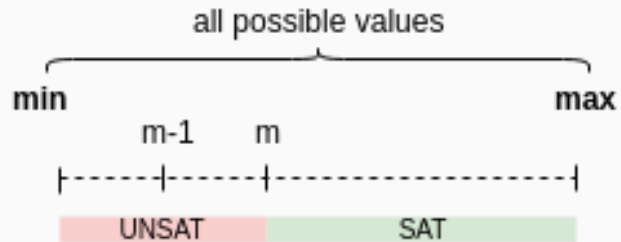
$$3x_2 + 4\overline{x_3} \leq 1,$$

$$3x_1 + 7\overline{x_3} \leq 1$$

$$3x_4 - 5x_5$$

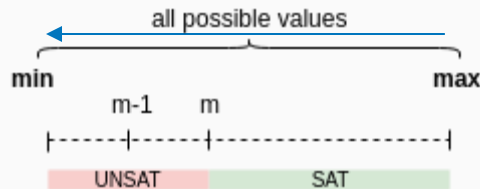
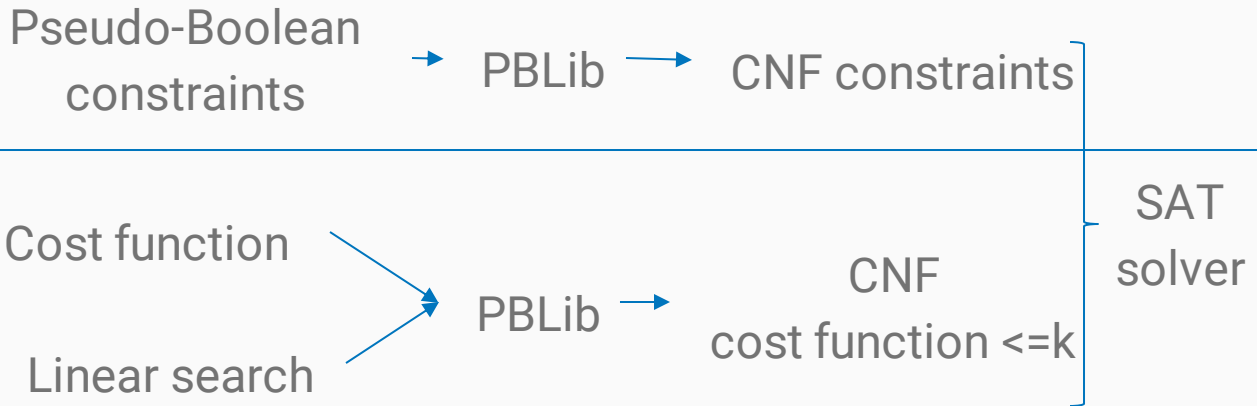
Search algorithms

Search space



Linear search

```
LinearSearchStrategy ls;  
PBDMin m = PBDMin(constraints, costFunction);  
Solver s(&ls,m);
```



Binary search

```
BinarySearchStrategy bs;  
PBMin m = PBMin(constraints, costFunction);  
Solver s(&bs,m);
```

Pseudo-Boolean
constraints



PBLib



CNF constraints

Cost function



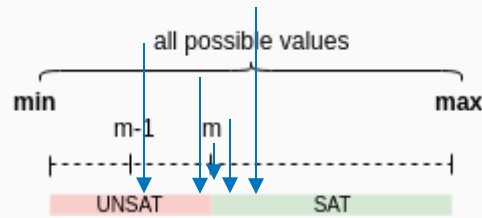
PBLib



CNF
cost function $\leq k$

Binary search

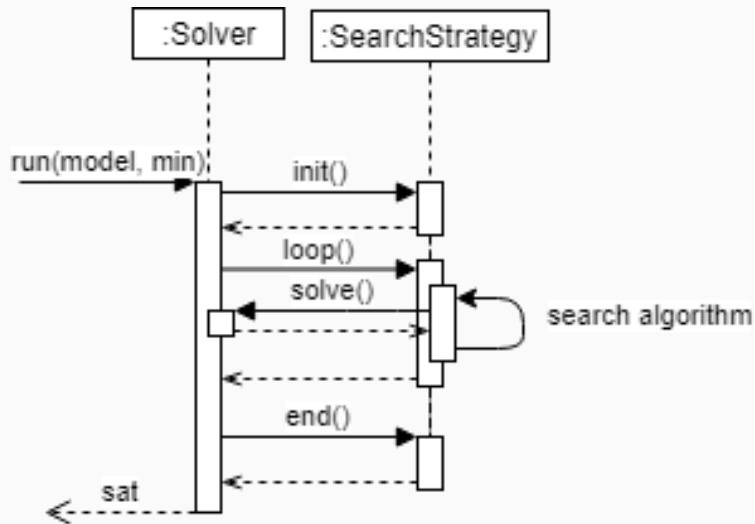
SAT
solver



Timeouts

Base solver

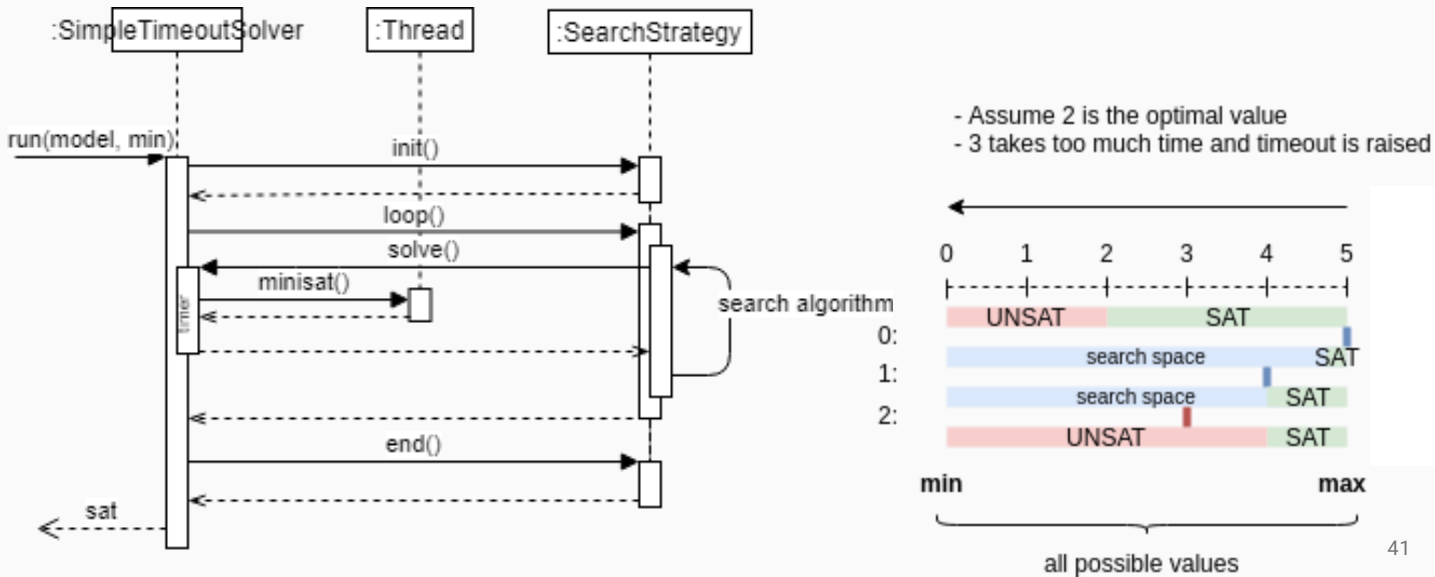
```
Solver solver(&search,problem);
```



Simple timeout

```
SimpleTimeoutSolver s(timeout,&search,problem);
```

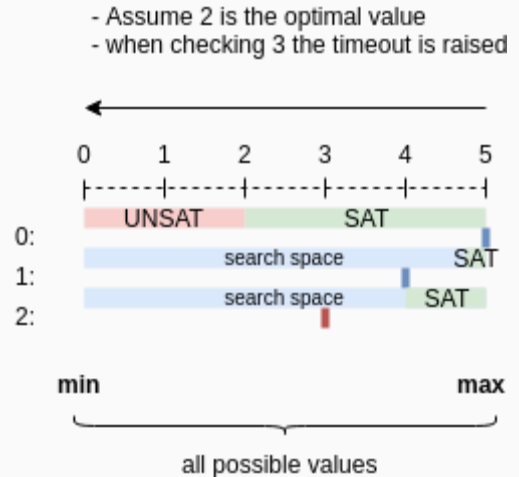
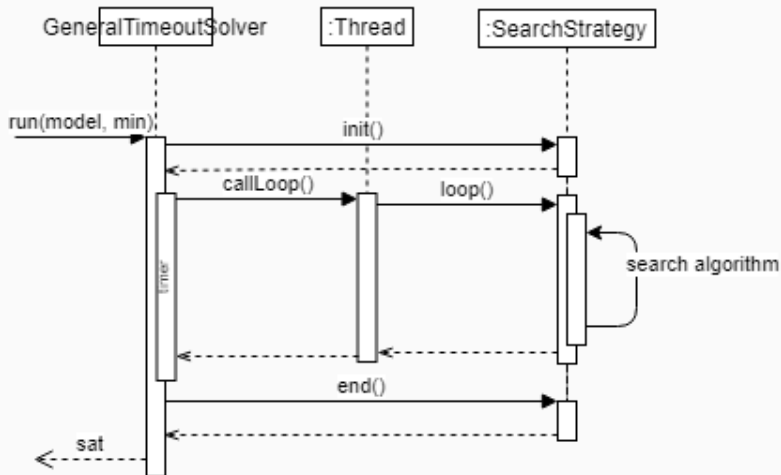
- Timeout for each call to the SAT Solver



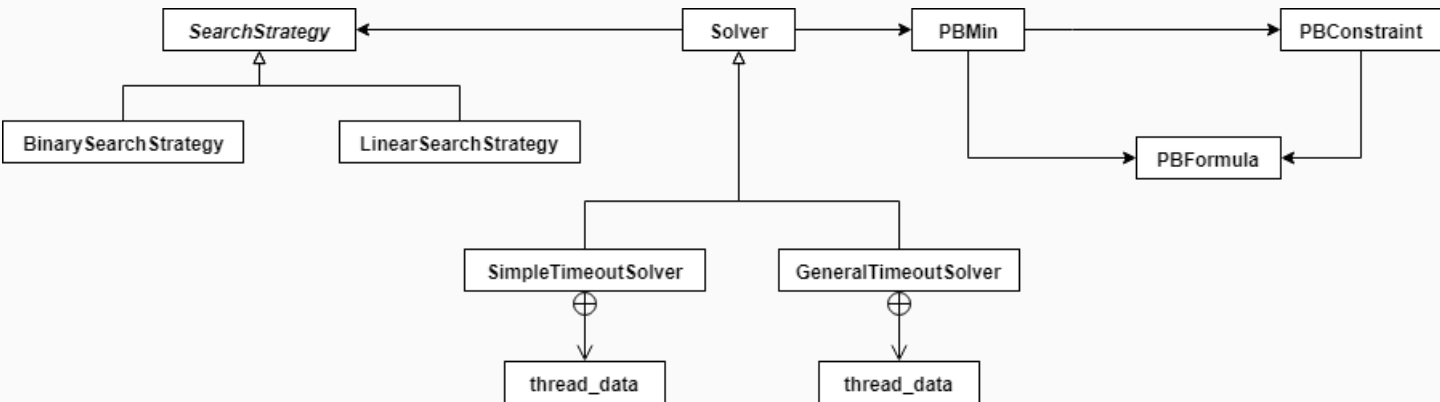
General timeout

```
GeneralTimeoutSolver s(timeout, &search, problem);
```

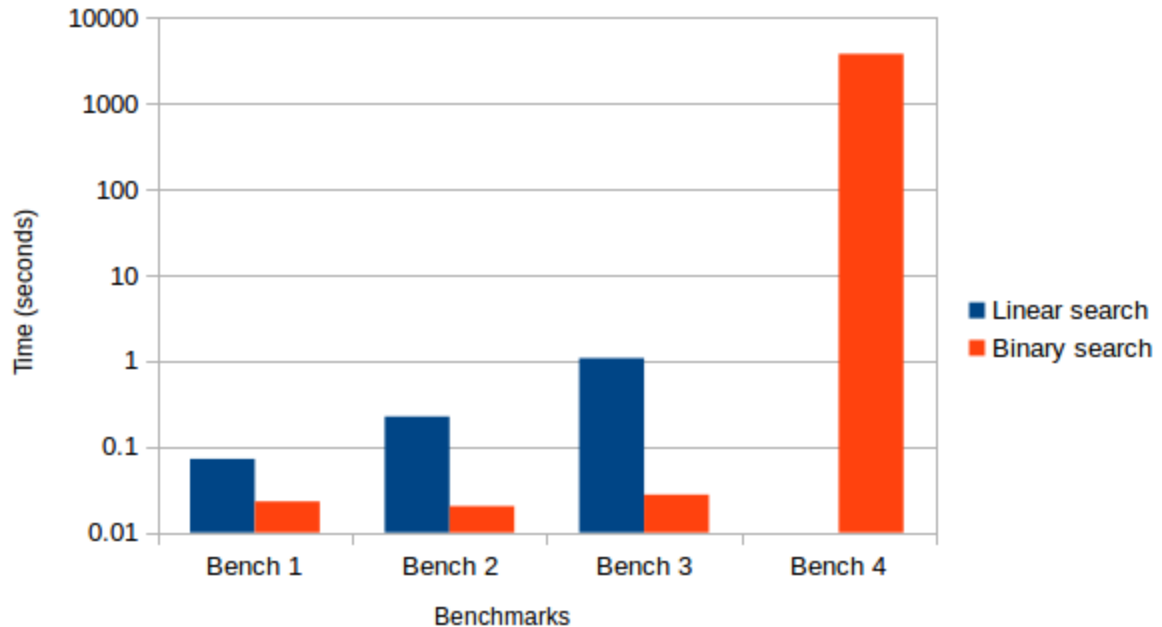
- Timeout for the whole problem



Architecture



Experiment



Conclusions & future work

Conclusions

- Goal of improving the overall time required to solve pseudo-Boolean minimisation problems
- A more user-friendly interface
- Two search algorithms to find the minimum value
- Two timeout strategies to finish the execution

Future work

- Github repository, documentation, installer and wiki
- Iteration 3: Multithreading
- Allow maximisation and other relational operators

Analysis of the planning

Timeline

Stage	Expected hours	Real hours
GEP	70	70
Requirement analysis, architecture and debugging	90	75
It 1: PB Minimisation	80	87
It 2: Timeout	80	88
It 3: Multithreading*	80	-
Finalization	50	65
Total	450	385

Budget

Estimated budget

Direct costs	10.015,00
Indirect costs	140,73
Contingency	3.025,61
Unforeseen	816,14
Total	13.997,48

Real budget

Direct costs	8.930,00
Indirect costs	257,73
Contingency	3.025,61
Unforeseen	816,14
Total	13.029,48

Sustainability

	PPP	Useful life	Risks
Environmental	7	20	-4
Economical	7	15	-10
Social	8	15	0
Sustainability range	58		

Thank you for
your time

Questions?



<https://github.com/marcbenedi/SAT-TFG>