

EFFICIENT MODEL BASED DIAGNOSIS WITH MAXIMUM SATISFIABILITY

Joao Marques-Silva^{1,2}, Mikoláš Janota¹, **Alexey Ignatiev**¹, and Antonio Morgado¹

July 31, 2015

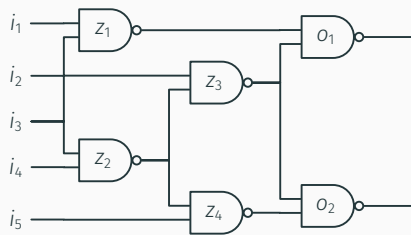
¹ INESC-ID, IST, University of Lisbon, Portugal

² CASL, University College Dublin, Ireland

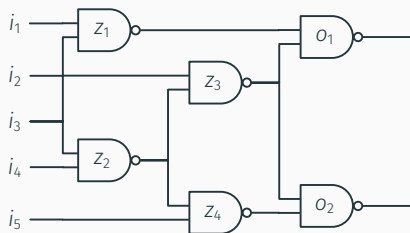
1. Model-based diagnosis
2. Dominator-oriented encoding
3. ITC99 benchmark instances
4. Experimental results
5. Summary and future work

MODEL-BASED DIAGNOSIS

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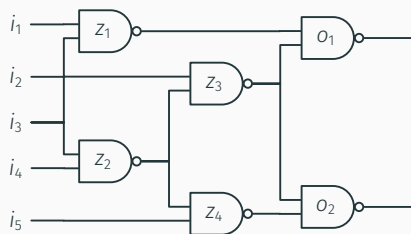
MODEL-BASED DIAGNOSIS



Comps \triangleq $\{z_1, z_2, z_3, z_4, o_1, o_2\}$

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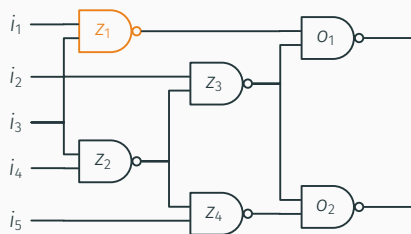
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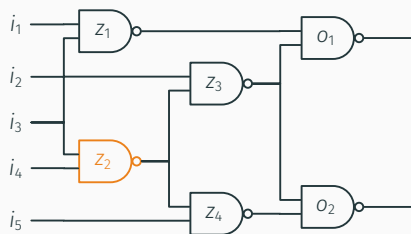
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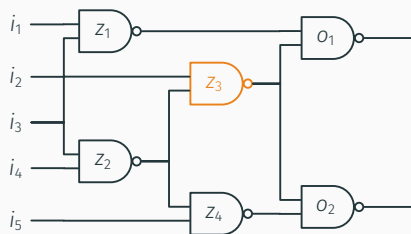
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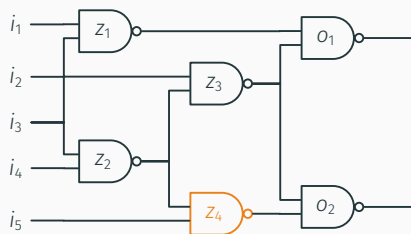
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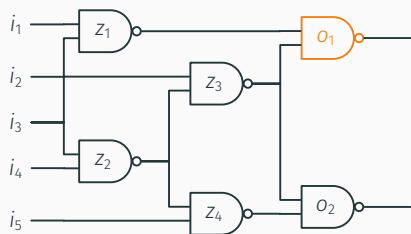
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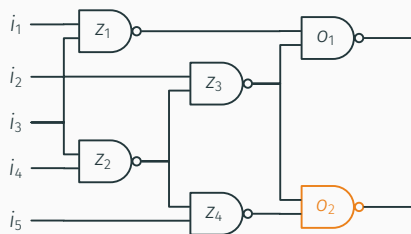
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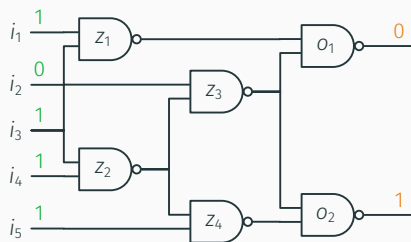
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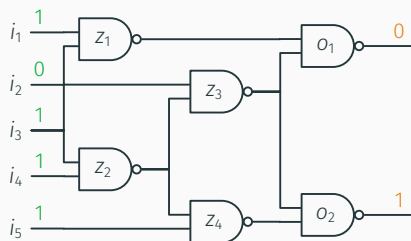
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Obs = $\{\langle i_1, i_2, i_3, i_4, i_5 \rangle = \langle 1, 0, 1, 1, 1 \rangle, \langle o_1, o_2 \rangle = \langle 0, 1 \rangle\}$



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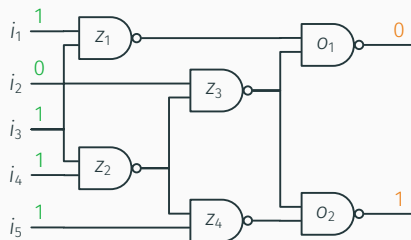
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$\text{SD} \wedge \text{Obs} \wedge \bigwedge_{c \in \text{Comps}} \neg \text{Ab}(c) \models \perp$

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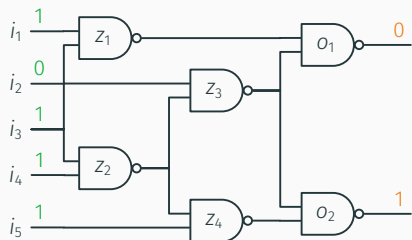
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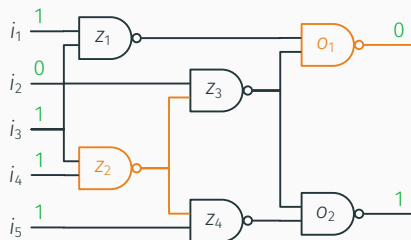
$\text{SD} \wedge \text{Obs} \wedge \bigwedge_{c \in \text{Comps}} \neg \text{Ab}(c) \neq \perp$



find $\Delta \subseteq \text{Comps}$ s.t.

$\text{SD} \wedge \text{Obs} \wedge \bigwedge_{c \in \Delta} \text{Ab}(c) \wedge \bigwedge_{c \in \text{Comps} \setminus \Delta} \neg \text{Ab}(c) \neq \perp$

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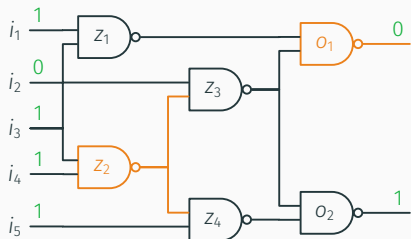


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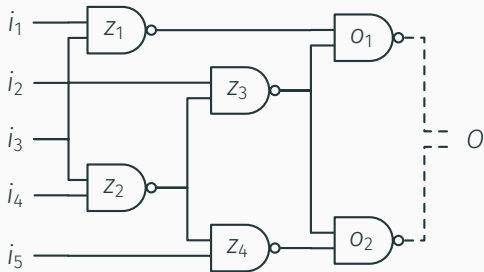
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$\Delta = \{z_2, o_1\}$ – minimize Δ with MaxSAT

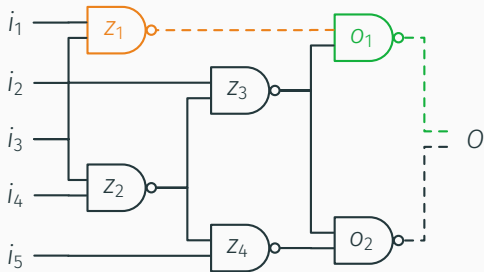
DOMINATOR-ORIENTED ENCODING

DOMINATORS AND TLDS



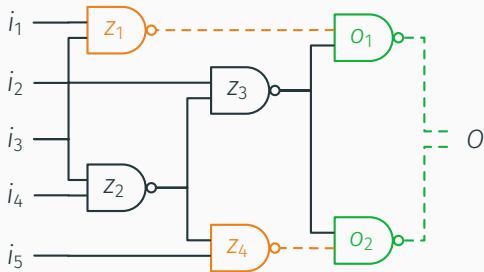
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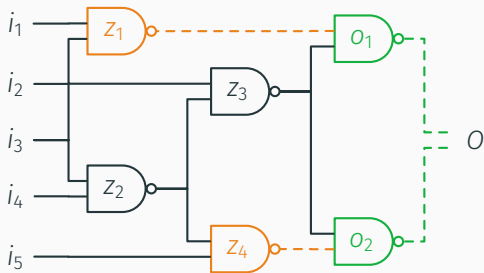
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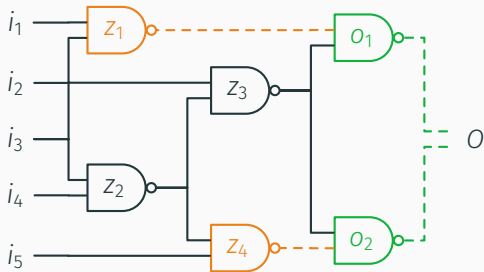
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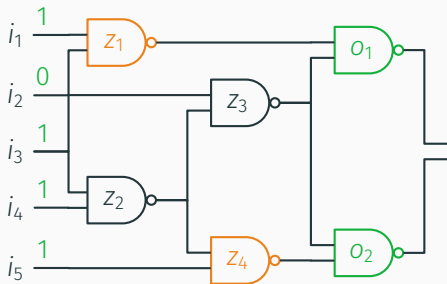
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dominated gates are hard

BACKBONE NODES

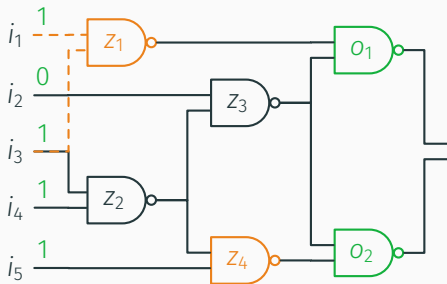


z_1 is *dominated* and its *output* is *fixed*



z_1 is a *backbone node*

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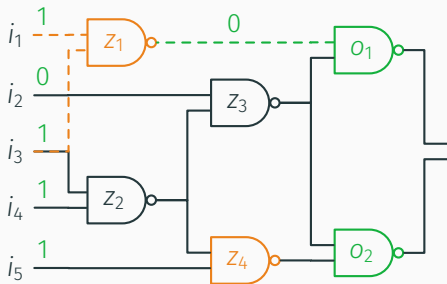


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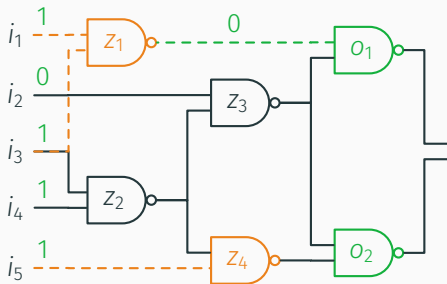


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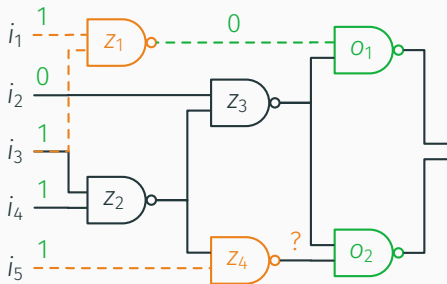


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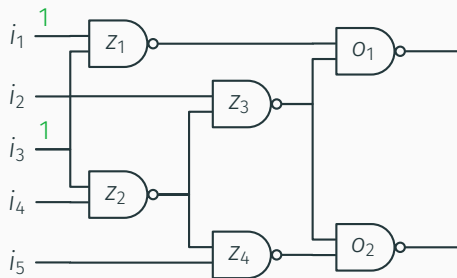


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BLOCKED EDGES

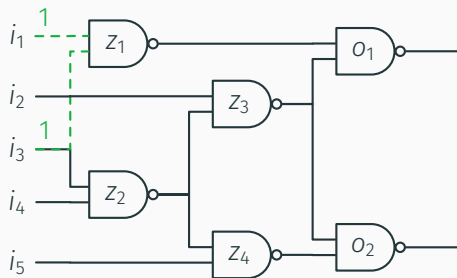


output of o_1 is *unchanged* for any value assigned to edge (z_3, o_1)



edge (z_3, o_1) is *blocked*

BLOCKED EDGES

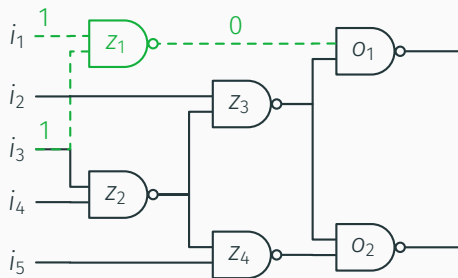


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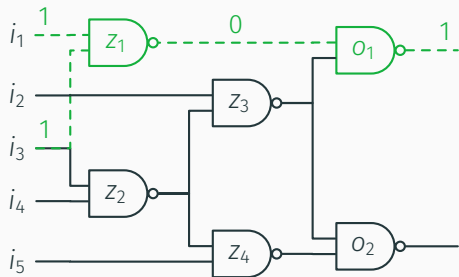


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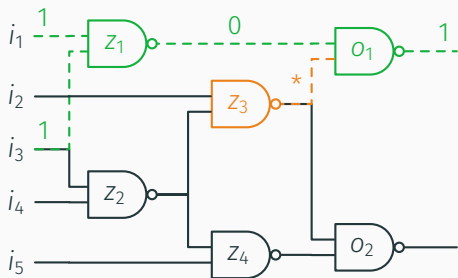


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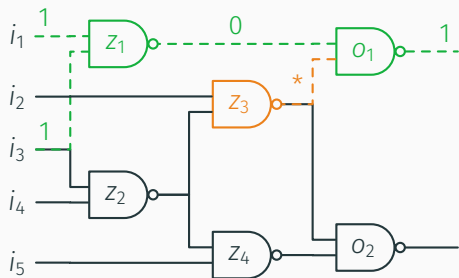


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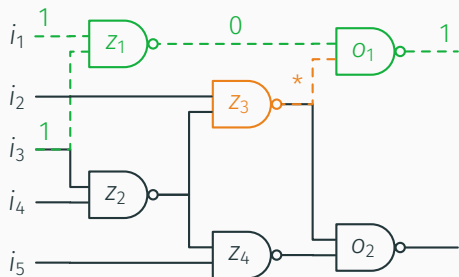


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FILTERED EDGES AND NODES



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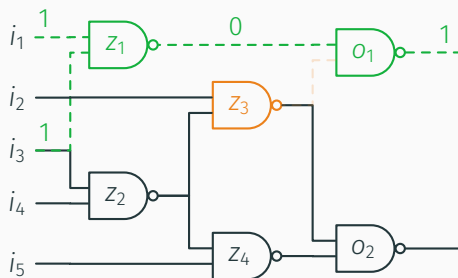


e is *blocked* or its *fanout node* is *filtered*



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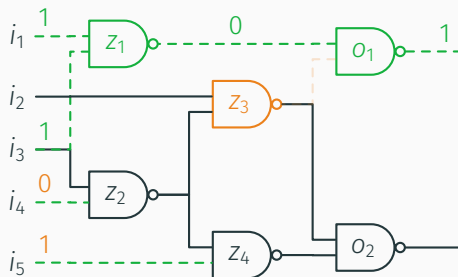


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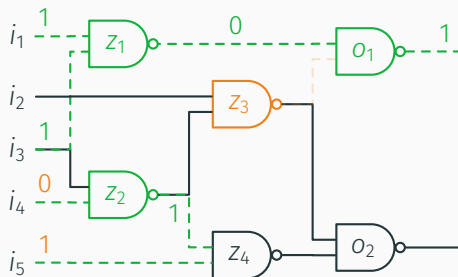


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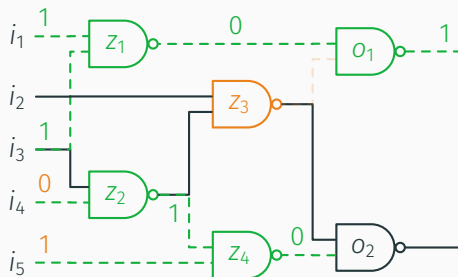


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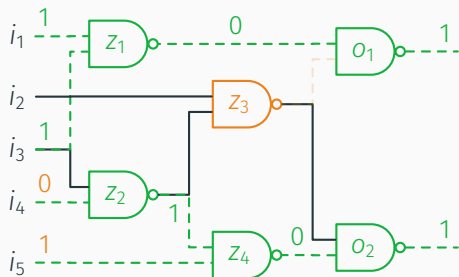


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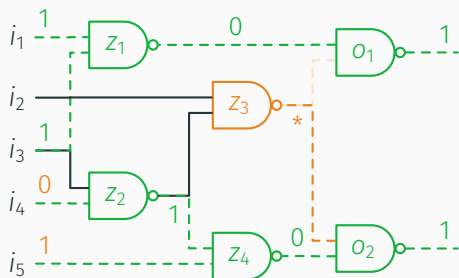


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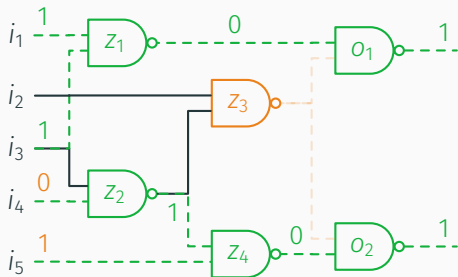


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FILTERED EDGES AND NODES



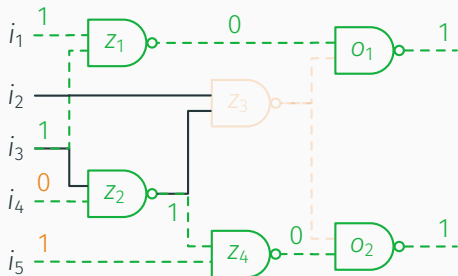
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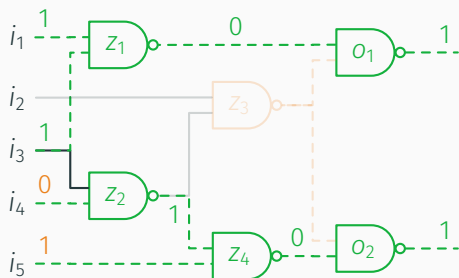
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```
1 global: ⟨SD, Comps, Obs⟩
2 repeat
3   | FindDominators()
4   | FindBackboneComponents()
5   | FindBlockedConnections()
6   | if MaxNumberIterations(): break
7 until NoMoreChanges()
8 GenMaxsatModel()
```

Algorithm 1: MBD to MaxSAT compilation

DOE approach:

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possible structural decompositions

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possible structural decompositions



subproblems can be solved separately

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2. iterate over k

But,

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core-guided MaxSAT can exploit this:

- $\text{Ab}(c) = 0 \quad \forall c \in \text{Comps}$ **by default**
- relax **on demand**, i.e. $\text{Ab}(c) = 1$ **when needed**

ITC99 BENCHMARK INSTANCES

ISCAS85 instances are *easy*

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new challenging suite — *ITC99*:

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- 8 ITC99 circuits

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- $\times 1000$ scenarios per circuit
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- -2097 instances (*hard for SATbD/SCryptoDiagnoser*) = 5903

EXPERIMENTAL RESULTS

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 1. ISCAS85 (16174 instances)
 2. ITC99 (5903 instances)

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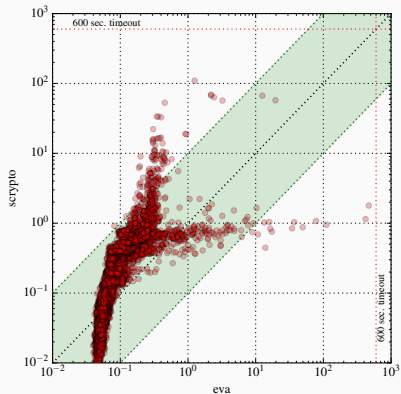
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- Machine configuration:
 - Intel Xeon E5-2630 2.60GHz with 64GByte RAM

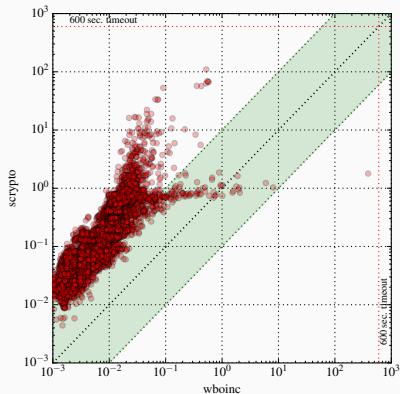
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 - 4GB memout



(a) eva vs. scripto



(b) wboinc vs. scripto

Figure 1: Scatter plots for ISCAS85 suite

16174 instances	scrypto	eva	wboinc
% solved	100.0	100.0	100.0
% scrypto wins	—	23.4	0.1
% eva wins	76.6	—	0.0
% wboinc wins	99.9	100.0	—

Table 1: Statistics for ISCAS85 suite

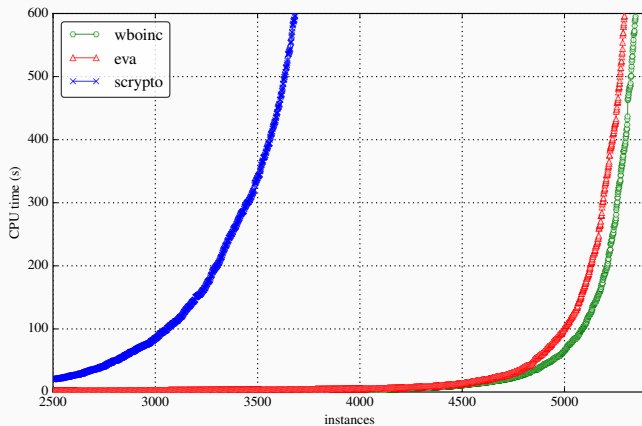


Figure 2: Cactus plot for ITC99 suite

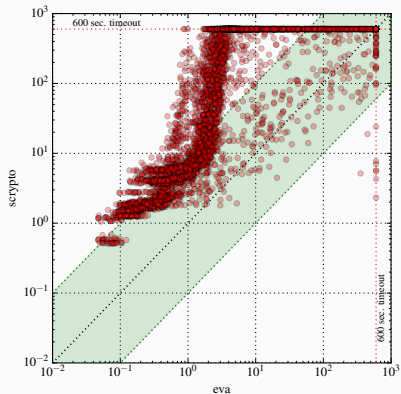
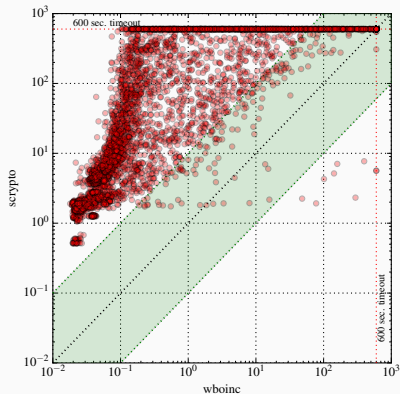
(a) *eva* vs. *scripto*(b) *wboinc* vs. *scripto*

Figure 3: Scatter plots for ITC99 suite

5903 instances	scrypto	eva	wboinc
% solved	62.4	89.7	90.6
% scrypto wins	—	2.2	0.4
% eva wins	97.8	—	13.4
% wboinc wins	99.6	86.6	—

Table 2: Statistics for ITC99 suite

SUMMARY AND FUTURE WORK

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- further optimizations for the DOE
- application to other related practical problems

QUESTIONS?