

Efficient Decision Ordering Techniques for SAT-based Test Generation

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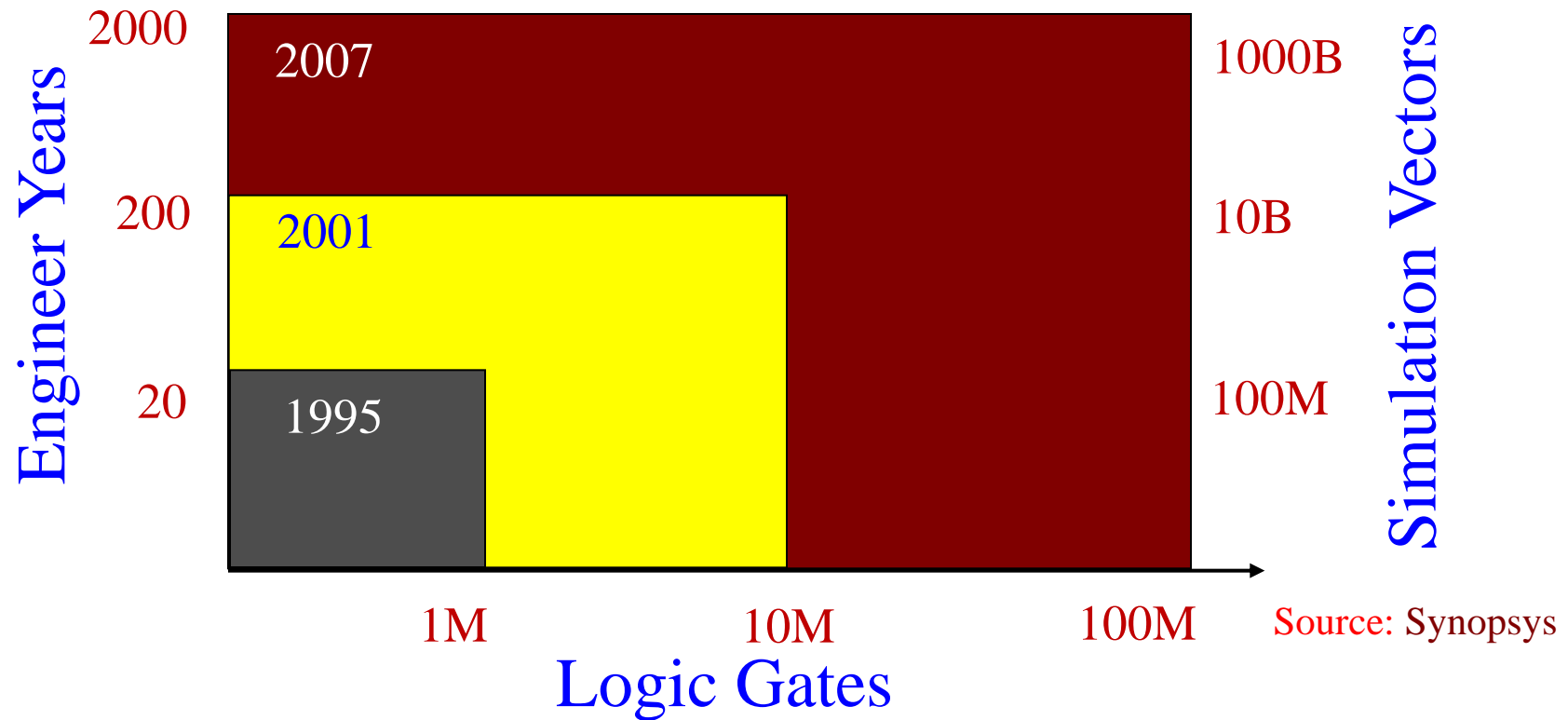
March 10, 2010



Outline

- Introduction
- Simulation-based Validation
 - ◆ Test Generation using Model Checking
 - ◆ Test Generation using SAT-based BMC
- Test Generation using Decision Ordering
 - ◆ Bit value ordering
 - ◆ Variable ordering
 - ◆ Test generation using our methodology
- Experiments
- Conclusion

Functional Verification of SOC Designs



- Functional validation is a major challenge
 - ◆ Majority of the SOC fails due to logic errors
- Simulation using directed tests is promising

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Test Generation using Model Checking

● Model Checking

- ◆ Design is modeled temporal specification, e.g., SMV
- ◆ Desired behaviors in temporal logic properties
- ◆ Property holds, or fails with a counterexample

Problem: Test generation is very costly or not possible in many scenarios in the presence of complex SoCs and/or complex properties.

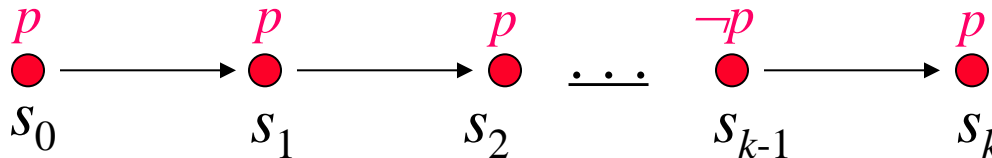
Approach: Exploit learning to reduce complexity

- Reduction of TG time & memory requirements
- Enables test generation in complex scenarios

SAT-based Bound Model Checking

- For every finite model and a LTL property ϕ there exists k such that: $M \models_k \phi \rightarrow M \models \phi$
- Test generation needs to consider **safety** properties
- The safety property P is valid up to cycle k iff $\Omega(k)$ is not satisfiable.

$$\Omega(k) = I(S_0) \wedge \bigwedge_{i=0}^{k-1} R(S_i, S_{i+1}) \wedge \bigvee_{i=0}^k \neg P(s_i)$$



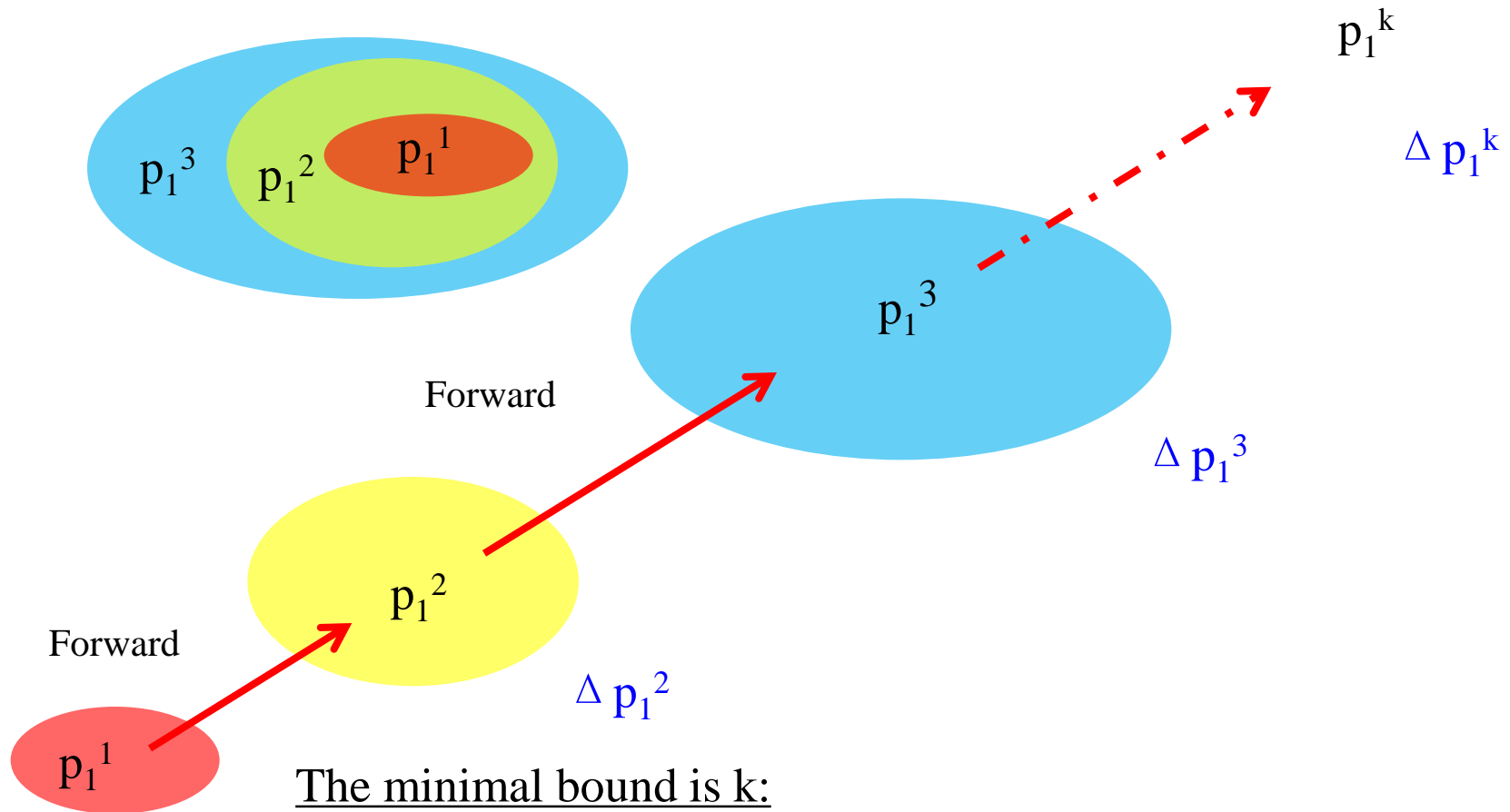
- If $\Omega(k)$ is satisfiable, then we can get an assignment which can be translated to a **test**.

DPLL Algorithm

```
while (1){  
    run_periodic_function();  
    if( decide_next_branch() ){  
        while ( Implication == CONFLICT) {  
            blevel = Conflict Backtrack ();  
            if( blevel<0 )  
                return UNSAT;  
        }  
    } else return SAT;  
} BCP = Implication Number + Conflict Backtrack
```

Boolean Constraint Propagation (BCP) consumes up to 80% of the time and resources during SAT solving

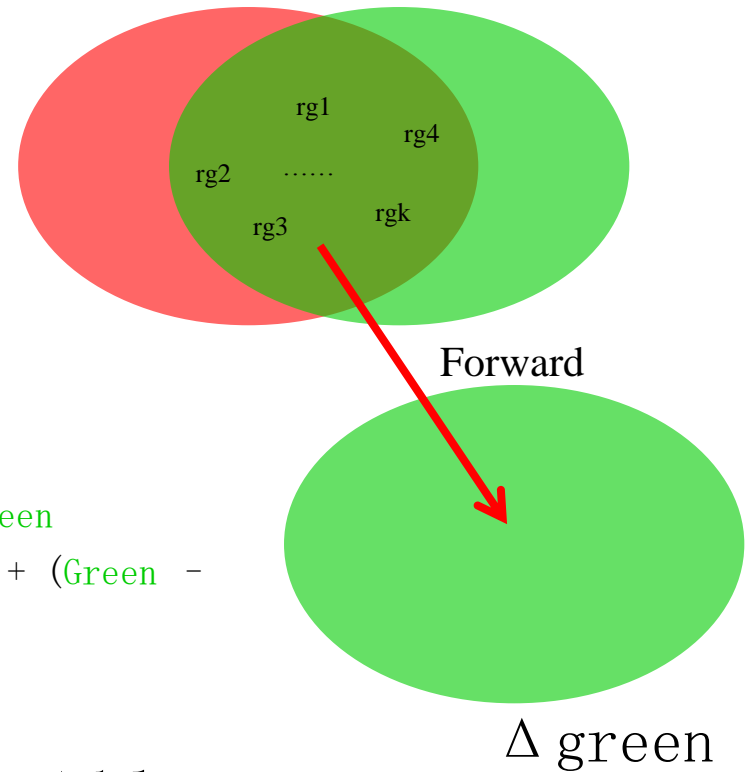
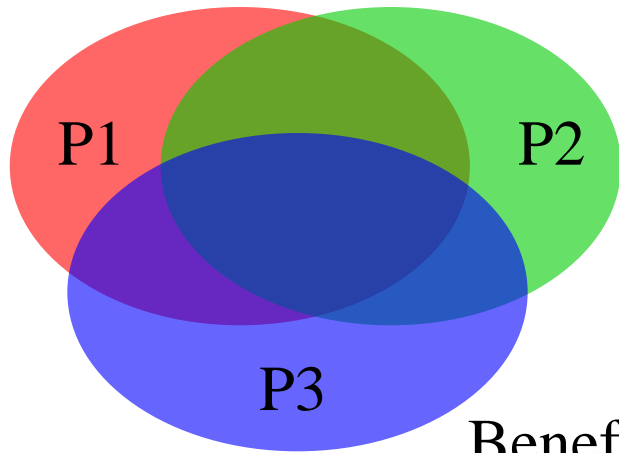
Same Property but Different Bounds



Save: $\Delta p_1^2 + \Delta p_1^3 + \dots + \Delta p_1^{k-1} + \dots + \Delta p_1^k$

O. Strichman. Pruning Techniques for the SAT-Based Bounded Model Checking Problems. CHARME, 2001

Same Design, Different Properties

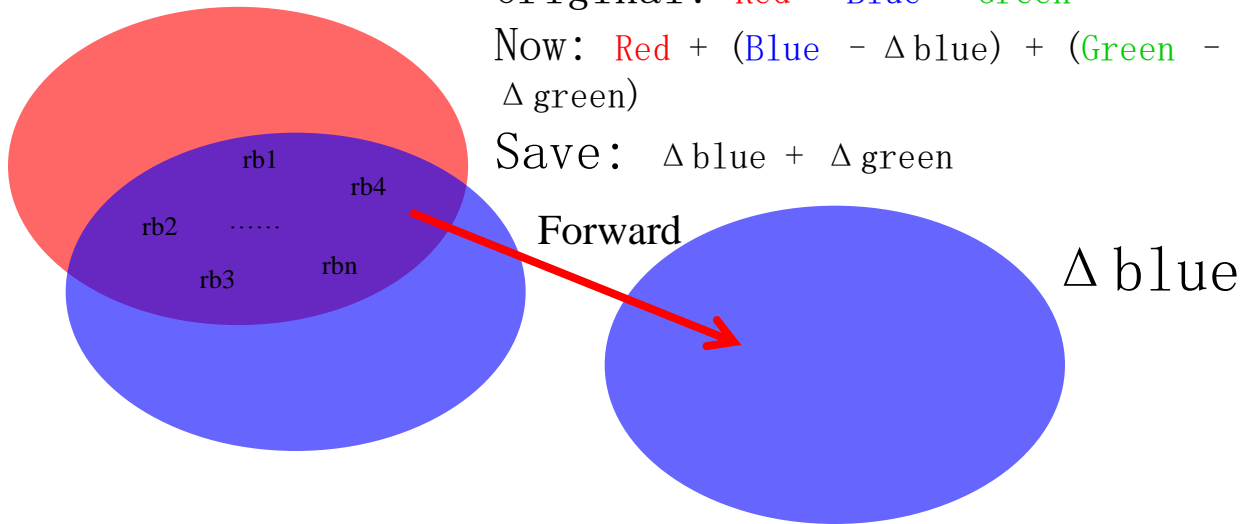


Benefit:

Original: Red + Blue + Green

Now: Red + (Blue - Δ blue) + (Green - Δ green)

Save: Δ blue + Δ green



Promising Observations

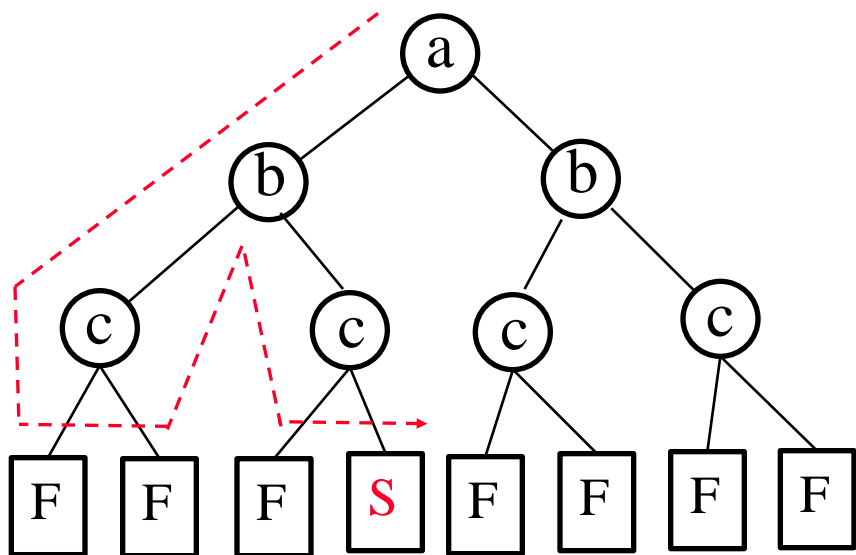
- Similar properties have the similar counter-examples (variable assignments).
 - ◆ Such important information can be reused.
- Current decision ordering techniques focus on the SAT problem instead of the real design.
 - ◆ For example, VSDIS, for each literal *lit* has a score
 - ◆ Initialization
 - ❖ ***score(lit) = literal count of lit in CNF clauses***
 - ◆ Periodical update (not include initialization)
 - ❖ ***score(lit) = score(lit) / 2 + lit_in_conflict(lit)***

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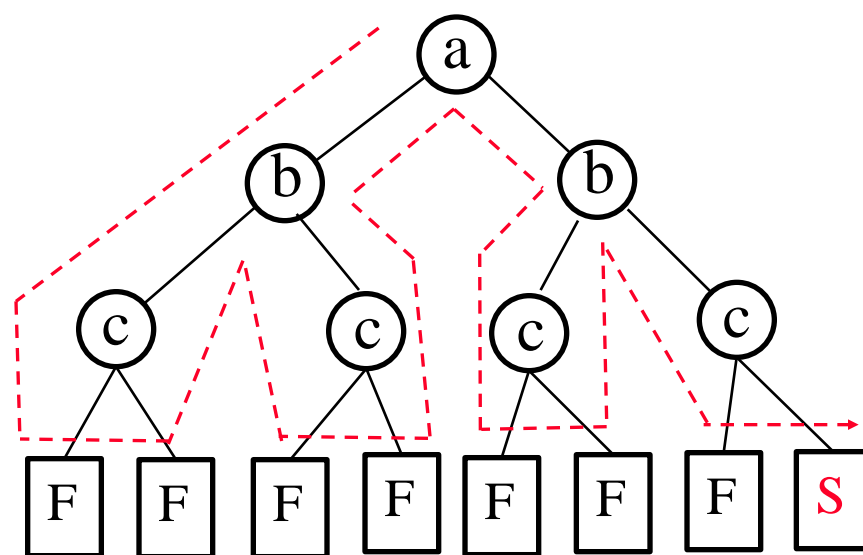
Two Similar SAT Problems

SAT 1



Ordering: a, a', b, b', c, c'

SAT 2



Ordering: a, a', b, b', c, c'

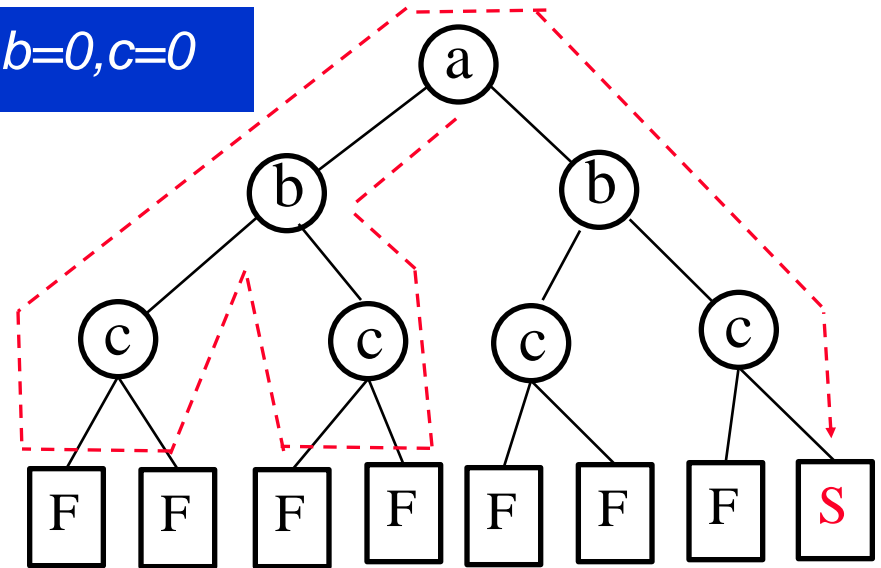
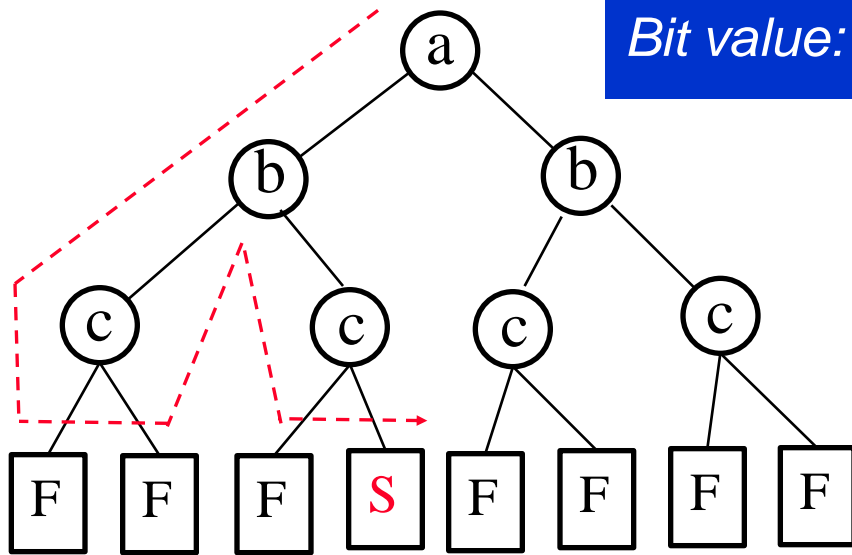
Without Learning, 7 conflicts in SAT2.

Learning: Bit Value Ordering

SAT 1

SAT 2

Bit value: $a=1, b=0, c=0$



Ordering: a, a', b, b', c, c'

Ordering: a, a', b', b, c', c

With bit value learning, 4 conflicts in SAT2.

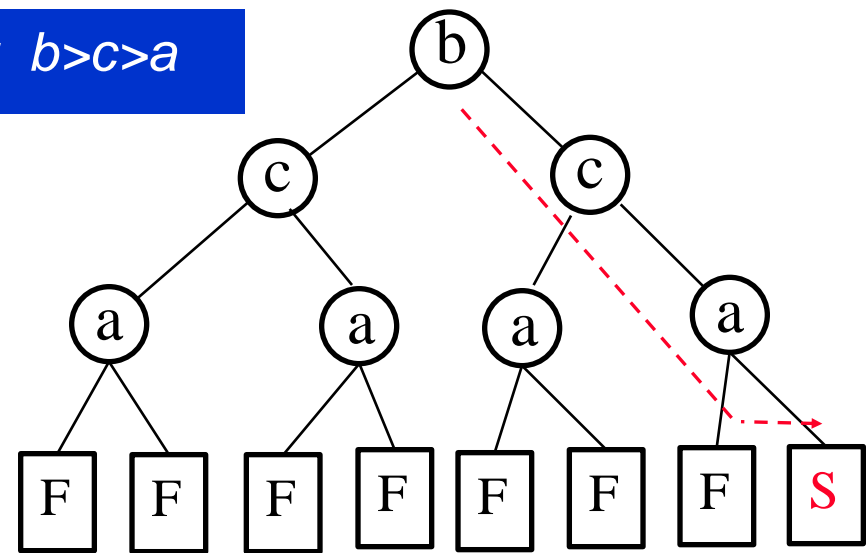
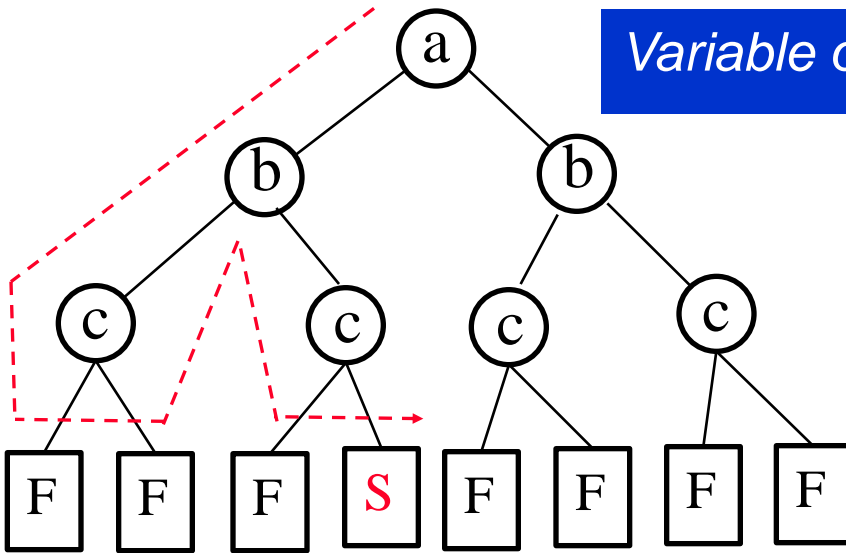
Learning: Variable Ordering

SAT 1

Bit value: $a=1, b=0, c=0$

Variable order: $b > c > a$

SAT 2



Ordering: a, a', b, b', c, c'

Ordering: b', b, c', c, a, a'

With bit value+ variable order learning, 1 conflict in SAT2.

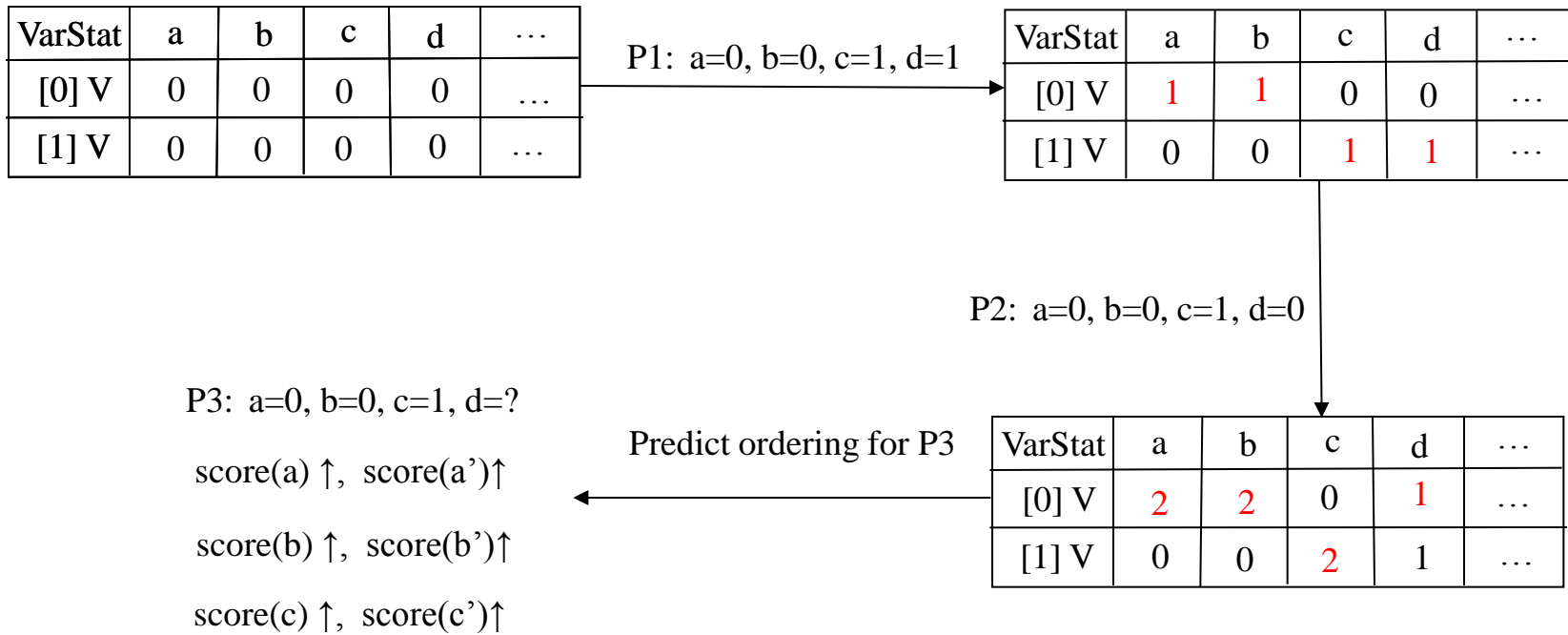
Test Generation Using Our Method

Inputs: a) Formal model, D

b) A cluster of properties P with satisfiable bounds

1. Initialize $varStat$
2. Select the base property p_1 , and generate CNF_1
3. $(assignment_1, test_1) = SAT(CNF_1)$
4. $Test-suite = \{test_1\}$
5. for i is from 2 to the size of P
 - a) Update $varStat$ using $assignment_{i-1}$
 - b) Generate $CNF_i = BMC(D, p_i, bound_i)$
 - c) $(assignment_i, test_i) = SAT(CNF_i)$
 - d) $Test-suite = Test-suite \cup \{test_i\}$endfor
6. Return $Test-suite$

An Illustrative Example with 3 properties



Approach: Using the statistics of the counterexamples when checking the properties in a cluster

- Count the number of values \rightarrow bit value ordering
- Variance of counts of two literals \rightarrow variable ordering

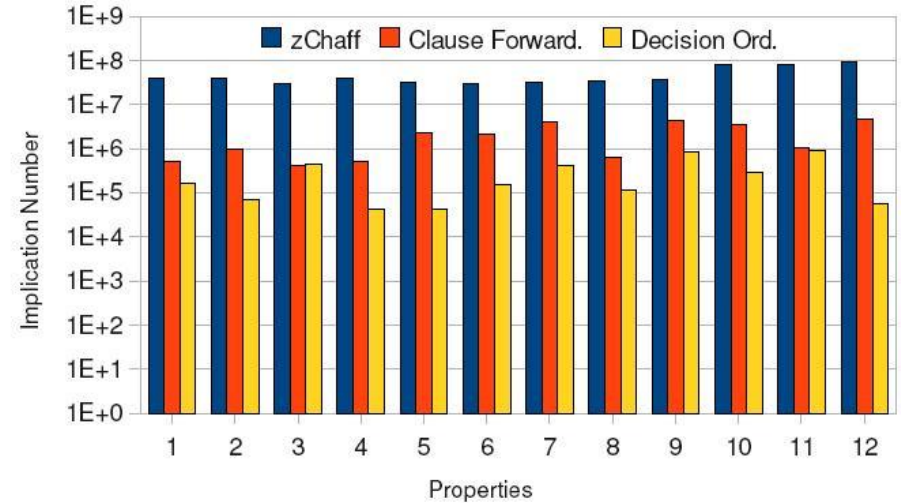
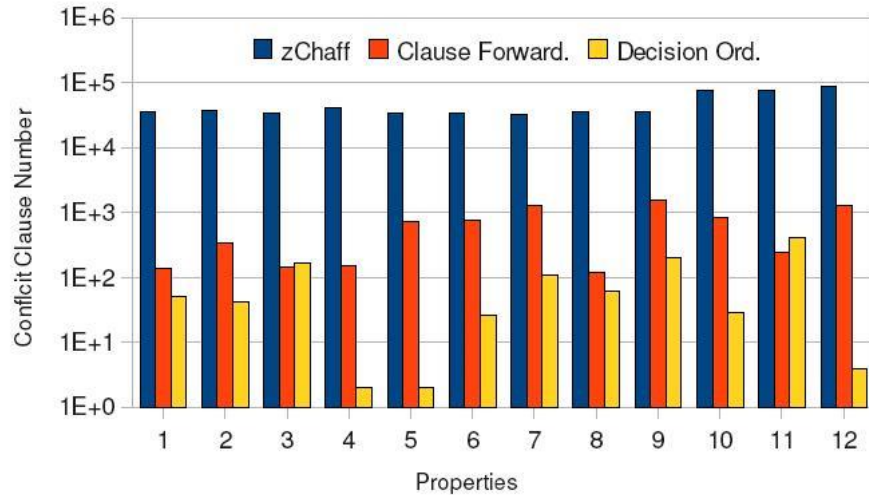
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Case Study: MIPS Processor

Property (test)	zChaff (sec)	Conflict Clause Forwarding	Improvement Factor	Decision Ordering	Improvement Factor
ALU	23.20	23.20	1	23.20	1
P1	20.73	2.74	7.57	0.18	15.22
P2	21.33	3.01	7.09	0.15	20.07
P3	18.03	2.70	6.68	0.29	9.31
DIV	18.78	18.78	1	18.78	1
P4	23.55	2.72	8.66	0.13	20.92
P5	18.31	3.60	5.09	0.14	25.71
P6	18.11	3.72	4.87	0.18	20.67
FADD	22.90	22.90	1	22.90	1
P7	16.95	4.46	3.80	0.23	19.39
P8	18.89	2.71	6.97	0.16	16.94
P9	19.80	4.70	4.21	0.39	12.05
MUL	64.21	64.21	1	64.21	1
P10	59.15	3.36	17.60	0.24	14.00
P11	59.65	3.85	15.49	0.45	8.56
P12	73.98	6.28	11.78	0.18	34.89

Case Study: MIPS Processor



Indications: Test generation complexity is significantly improved

- Reduction of conflict clauses
- Reduction of implication number

Case Study: OSES

- This case study is a on-line stock exchange system. The activity diagram consists of **27** activities, **29** transitions and **18** key paths.

Cluster	Size	zChaff	Conflict Forward	Improve ment Factor	Decision Ordering	Improvement Factor
C1	3	1.18	2.18	0.54	0.70	3.11
C2	4	14.53	9.53	1.52	0.78	12.22
C3	8	375.91	170.06	2.21	36.19	4.70
C4	4	12.98	8.33	1.56	1.24	6.72
C5	4	7.13	16.88	0.42	1.02	16.55
C6	8	720.13	474.68	1.52	28.60	16.60
C7	4	10.80	24.55	0.44	1.95	12.59
C8	8	656.95	321.14	2.05	77.65	4.14
C9	8	248.17	82.42	3.01	37.93	2.17
Average	-	227.53	123.21	1.85	20.67	5.97

Conclusions

- Functional validation is a major bottleneck
 - ◆ SAT-based approaches are promising for automated test generation.
- Proposed an efficient technique for generation of directed tests using learning techniques
 - ◆ Developed a novel decision ordering technique using both bit-value ordering and variable ordering
- Successfully applied on both hardware and software designs
 - ◆ Significant reduction in overall validation effort



Thank you !