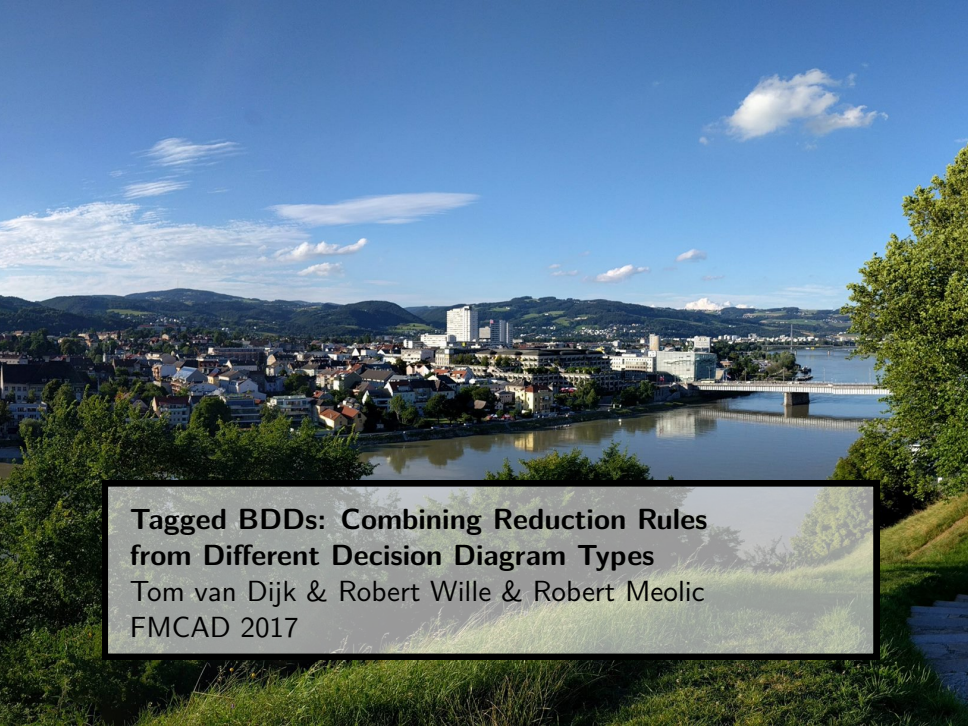




Tagged BDDs: Combining Reduction Rules from Different Decision Diagram Types

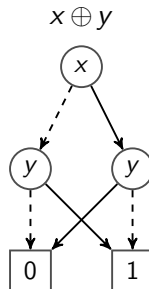
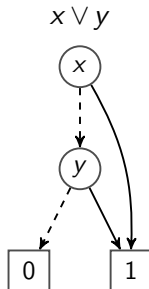
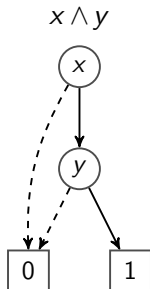
Tom van Dijk & Robert Wille & Robert Meolic
FMCAD 2017

TL; DR

- ▶ **Normal BDDs:**
Great with symmetries and dont-cares
- ▶ **Zero-suppressed BDDs:**
Great with subsets and sparse matrices
- ▶ **Tagged BDDs:**
Combining both types to use both features simultaneously
TBDDs implemented in the BDD package “Sylvan”
- ▶ **Application:**
On-the-fly symbolic learning of transition systems (LTSmin)

Binary Decision Diagrams

- ▶ A BDD is a **directed acyclic graph** encoding a $\mathbb{B}^k \rightarrow \mathbb{B}$ function or a $S \subseteq \mathbb{B}^k$ set
- ▶ Paths represent valuations of $\mathbb{B}^k$
- ▶ For sets: path to 1, then valuation is in the set

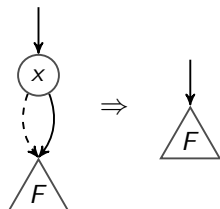


BDDs and ZBDDs

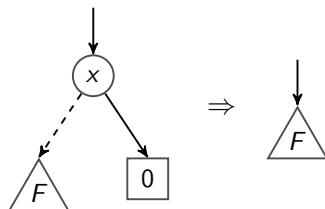
Zero-suppressed BDDs

- ▶ Different “reduction rule”
 - ▶ BDD: skipped variables are dont-cares
 - ▶ ZBDD: skipped variables are set to false
- ▶ Different applications: sparse matrices, subsets

BDD reduction rule



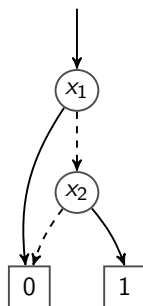
ZBDD reduction rule



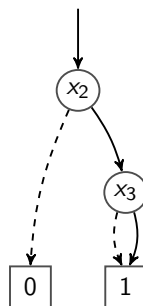
BDDs and ZBDDs

BDDs and ZBDDs

Two decision diagrams representing $f(x_1, x_2, x_3) = \overline{x_1}x_2$, i.e., the set $\{\overline{x_1}x_2x_3, \overline{x_1}x_2\overline{x_3}\}$



(a) BDD



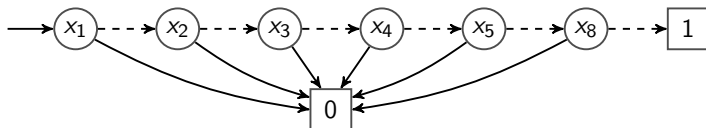
(b) ZBDD

Example

- ▶ A set on variables x_1 to x_8
- ▶ Four items: $\{00000000, 00000010, 00000100, 00000110\}$.

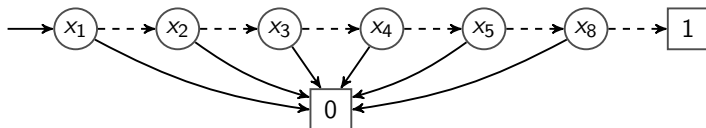
Example

- ▶ A set on variables x_1 to x_8
- ▶ Four items: $\{00000000, 00000010, 00000100, 00000110\}$.
- ▶ As a normal BDD

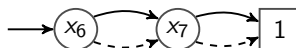


Example

- ▶ A set on variables x_1 to x_8
- ▶ Four items: $\{00000000, 00000010, 00000100, 00000110\}$.
- ▶ As a normal BDD



- ▶ As a Zero-suppressed BDD



Core idea

- ▶ Apply both types of reduction
 - ▶ Rule 1: skip nodes with identical edges
 - ▶ Rule 2: skip nodes with true edge to false
- ▶ How to distinguish which rule was applied?

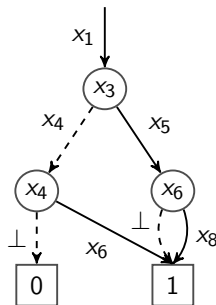
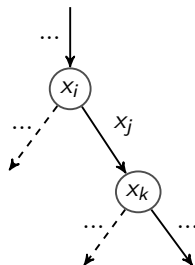
Core idea

- ▶ Apply both types of reduction
 - ▶ Rule 1: skip nodes with identical edges
 - ▶ Rule 2: skip nodes with true edge to false
- ▶ How to distinguish which rule was applied?
- ▶ Solution: a variable label (“tag”) on every edge
 - ▶ Missing nodes $x_i < x_{\text{tag}}$ due to rule 1
 - ▶ Missing nodes $x_i \geq x_{\text{tag}}$ due to rule 2
- ▶ Maximally apply both rules!

Tagged BDDs

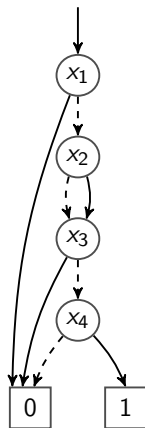
Examples

- ▶ We expect nodes between x_i and x_k
- ▶ Missing nodes $x_i < x < x_j$ due to rule 1 (BDD)
- ▶ Missing nodes $x_j \leq x < x_k$ due to rule 2 (ZBDD)



Alternating sequences

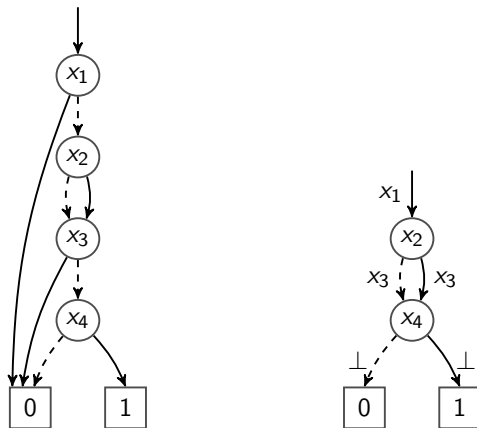
What if one variable label is not enough?



Tagged BDDs

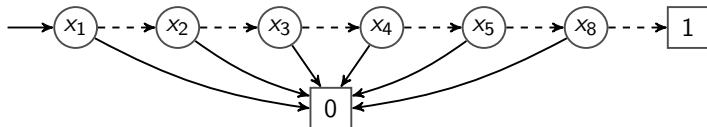
Alternating sequences

What if one variable label is not enough?

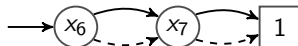


BDDs and ZBDDs

- ▶ A set on variables x_1 to x_8
- ▶ Four items: $\{00000000, 00000010, 00000100, 00000110\}$.
- ▶ As a normal BDD

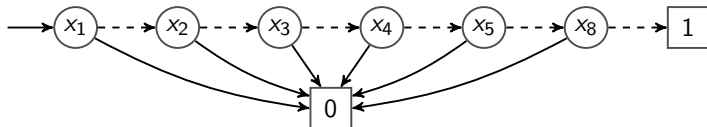


- ▶ As a Zero-suppressed BDD

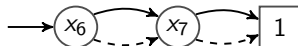


BDDs and ZBDDs

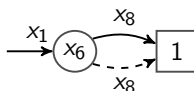
- ▶ A set on variables x_1 to x_8
- ▶ Four items: $\{00000000, 00000010, 00000100, 00000110\}$.
- ▶ As a normal BDD



- ▶ As a Zero-suppressed BDD

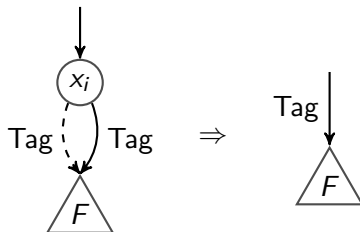


- ▶ As a Tagged BDD



Bottom-up reduction rules

x_i then $F(\text{Tag})$ else $F(\text{Tag})$

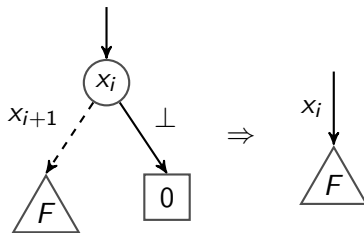


TBDD rules

Bottom-up reduction rules

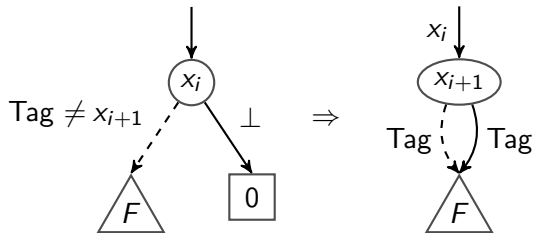
Note: x_{i+1} is $\perp$ if x_i is the last variable.

x_i then 0 else $F(x_{i+1})$

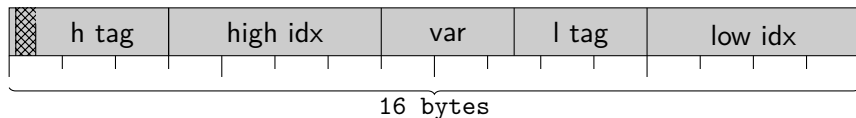


Bottom-up reduction rules

x_i then 0 else $F(\text{Tag})$



TBDD node layout



Each TBDD node is 16 bytes

- ▶ tags and variables: 20 bits
- ▶ edge indices: 32 bits
- ▶ extra bit for complementing (low edge)

Implementation

- ▶ Manipulating the 16-byte TBDD nodes
- ▶ Primitive for making nodes (`tbdd_makenode`)
 - ▶ This implements the three rules
- ▶ Operations
 - ▶ Binary operators
 - ▶ Negation (not trivial like BDDs)
 - ▶ Function domain extension (not trivial like BDDs)
 - ▶ Abstraction (exists, forall)
 - ▶ Relational product (for transition systems)
 - ▶ Other functions (counting, etc)
- ▶ Garbage collection and parallelism boilerplate (framework)

Outline

On-the-fly symbolic transition learning

- ▶ Learn a model symbolically starting from an initial state
- ▶ Interleave reachability with transition learning of new states
- ▶ Encode obtained new transitions in BDD
- ▶ Number of variables per state unknown (init to 0)
- ▶ BDDs and ZBDDs should both be strong

On-the-fly transition learning

- ▶ Compute reflexive transitive closure of $\mathcal{T}$ applied to $\mathcal{S}$
- ▶ Use a **frontier set** (only new states)

```
def ClosureFS( $\mathcal{S}, \mathcal{T}$ ):  
    states  $\leftarrow \mathcal{S}$   
    frontier  $\leftarrow \mathcal{S}$   
    while frontier  $\neq \emptyset$ :  
        next  $\leftarrow \text{RelProd}(\text{frontier}, \mathcal{T})$   
        frontier  $\leftarrow \text{next} \setminus \text{states}$   
        states  $\leftarrow \text{states} \cup \text{frontier}$   
    return states
```

On-the-fly transition learning

- ▶ Compute reflexive transitive closure of $\mathcal{T}$ applied to $\mathcal{S}$
- ▶ Use a frontier set (only new states)
- ▶ On-the-fly learning via (explicit) next-state interface

```
def ClosureFS( $\mathcal{S}$ ):  
    states  $\leftarrow \mathcal{S}$   
    frontier  $\leftarrow \mathcal{S}$   
     $\mathcal{T} \leftarrow \emptyset$   
    while frontier  $\neq \emptyset$ :  
         $\mathcal{T} \leftarrow \text{Learn}(\text{frontier}, \mathcal{T})$   
        next  $\leftarrow \text{RelProd}(\text{frontier}, \mathcal{T})$   
        frontier  $\leftarrow \text{next} \setminus \text{states}$   
        states  $\leftarrow \text{states} \cup \text{frontier}$   
    return states,  $\mathcal{T}$ 
```


Outline

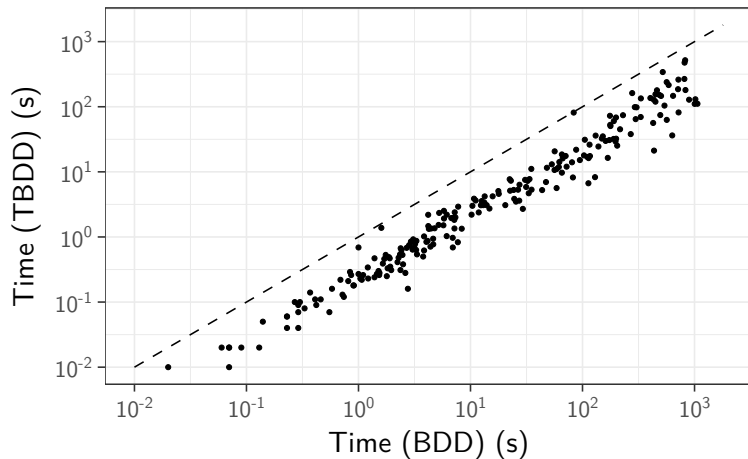
Results

- ▶ Implemented in Sylvan (parallel BDD package)
- ▶ Using LTSmin on the BEEM model database
- ▶ 48-core machine (4 processors $\times$ 12 cores)
- ▶ Publicly available: <http://fmv.jku.at/tbdd>

	BDD	TBDD
Time 1 core	24504 sec.	6453 sec.
Time 48 cores	14672 sec.	1075 sec.
#Nodes in the visited set	59,503,837	5,922,973

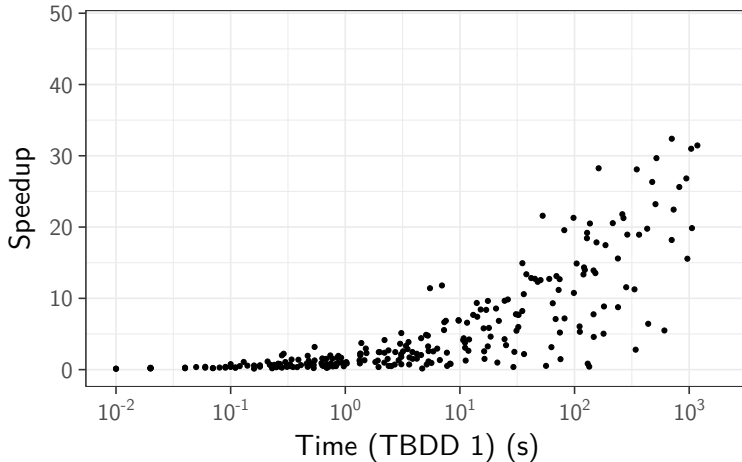
Experimental evaluation

Results: BDD vs TBDD (1 core)



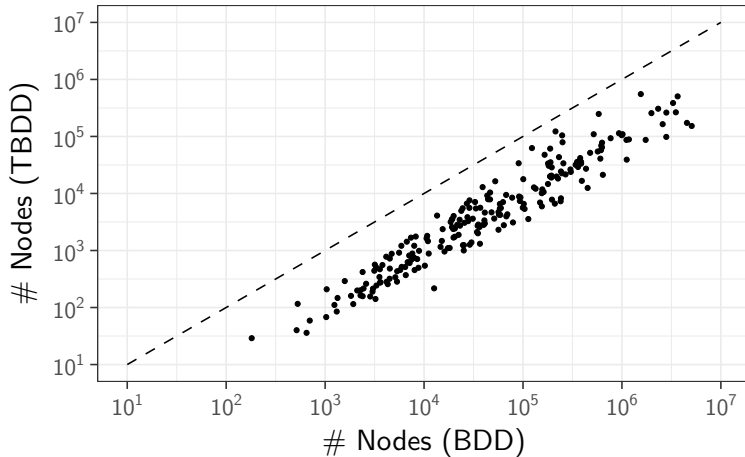
Experimental evaluation

Results: TBDD speedup



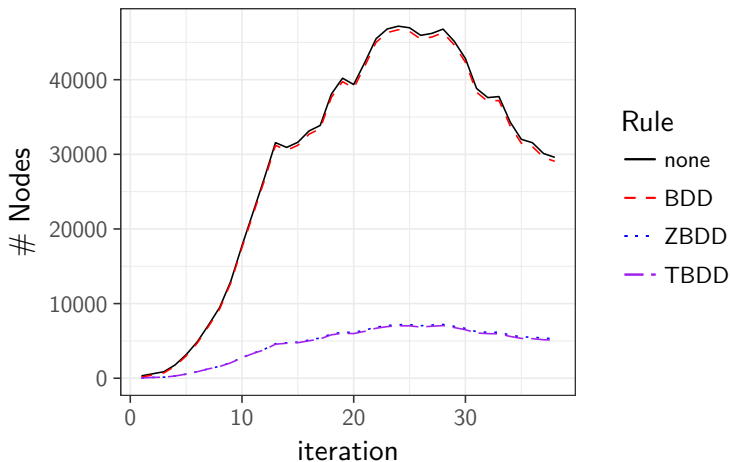
Experimental evaluation

Results: Number of nodes



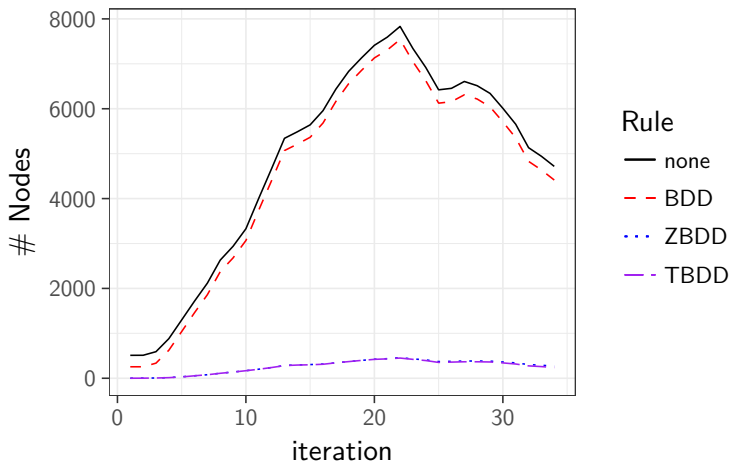
Experimental evaluation

Set of visited states of at.1



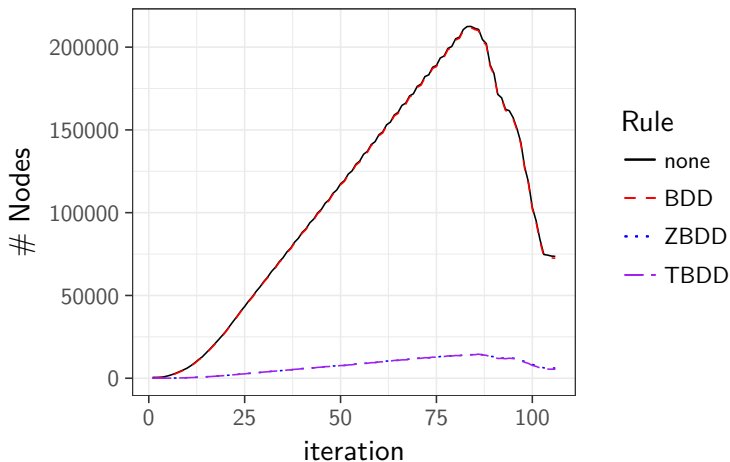
Experimental evaluation

Set of visited states of protocols.3.visited



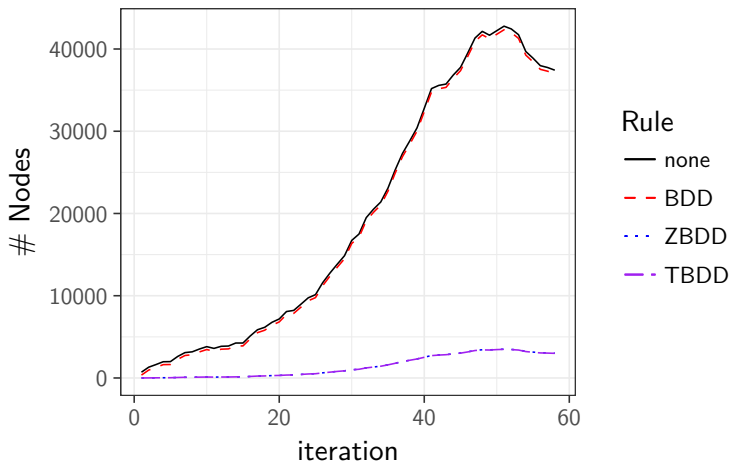
Experimental evaluation

Set of visited states of protocols.5.visited



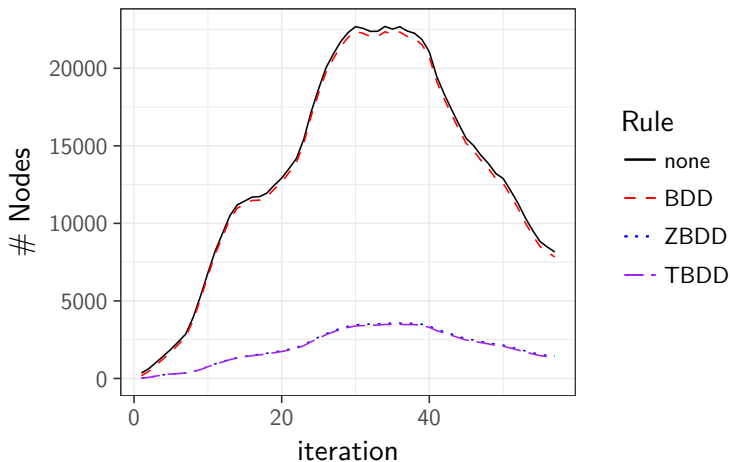
Experimental evaluation

Set of visited states of `lifts.1`



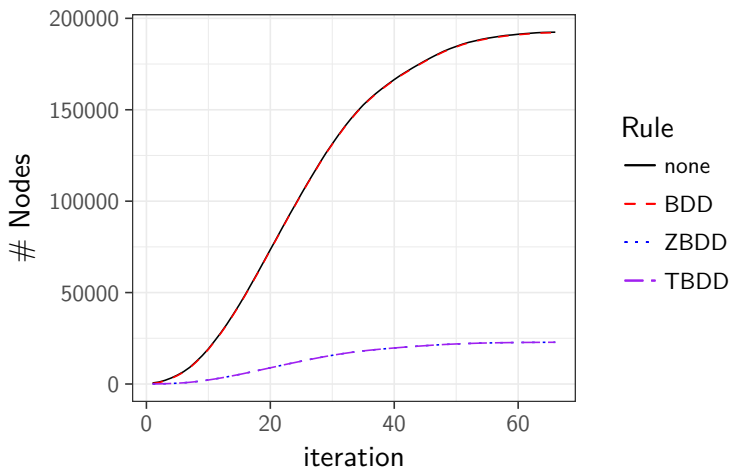
Experimental evaluation

Set of visited states of `lamport.1`



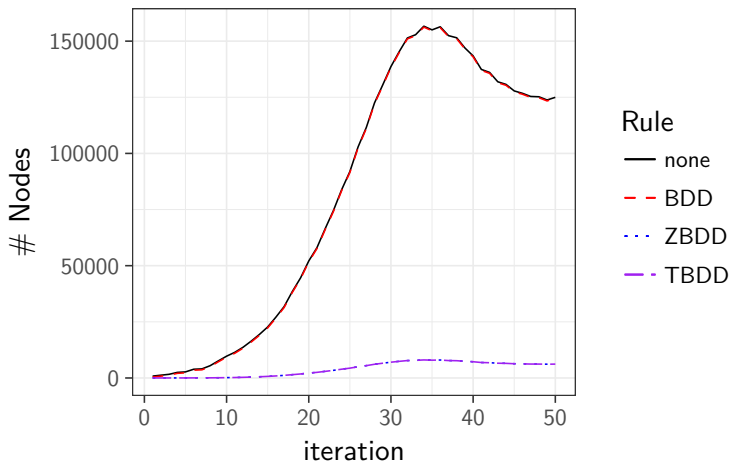
Experimental evaluation

Set of visited states of krebs .1



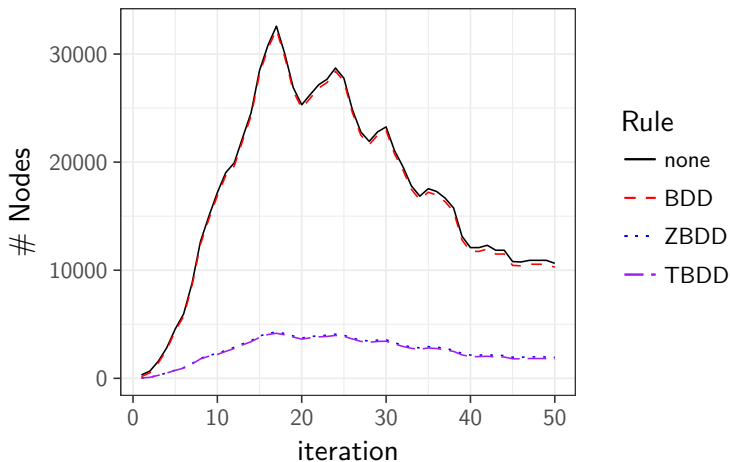
Experimental evaluation

Set of visited states of exit.1



Experimental evaluation

Set of visited states of `fischer.2`



Outline

Contributions and Conclusions

- ▶ Tagged BDDs to combine BDD and ZBDD rules.
- ▶ Implemented in Sylvan with multi-core implementation
<https://www.github.com/trolando/sylvan>
- ▶ Faster than normal BDDs on BEEM models
- ▶ Good parallel performance
- ▶ BEEM models disappointing w.r.t. BDD rule

Future Directions

- ▶ Apply to better symbolic models, e.g., Petri nets?
- ▶ Study other rules and more complicated rules
- ▶ Dynamic variable reordering and other operations

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