# Accelerating Predicate Abstraction for Probabilistic Automata

Dimitri Bohlender

RWTH Aachen University

September 12, 2014 / Master Thesis Presentation

Outline

Motivation

Why Model Checking?



Outline

Motivation

# Why Model Checking?

• testing cannot prove absence of bugs



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- testing cannot prove absence of bugs
- formal proof

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#### **Properties**

eventually a collision free transmission occurs



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### Why Probabilistic Model Checking?

#### **Properties**

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#### Why Probabilistic Model Checking?

Verification of probabilistic models, e.g. network protocols

#### **Properties**

- eventually a collision free transmission occurs
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### Why Model Checking?

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# Why Probabilistic Model Checking?

Verification of probabilistic models, e.g. network protocols

#### **Properties**

- eventually a collision free transmission occurs
- no collision ever occurs
- $\bullet$  probability for a collision is below 5%

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Outline

Motivation

# State Space Explosion

Even "simple" system descriptions yield huge state spaces



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### State Space Explosion

Even "simple" system descriptions yield huge state spaces

⇒ construction & analysis often infeasible (memory & time constraints)

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#### Observation

Model often more detailed than necessary to check property of interest

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#### **Approaches**



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analyse over-approximating, abstract model instead

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#### **Approaches**

- analyse over-approximating, abstract model instead (Menu-game)
- use space-efficient, symbolic data structures (BDD)

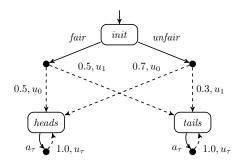
Optimisations

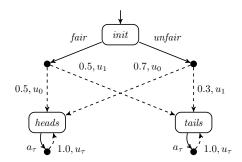
- 1 Probabilistic Models and Symbolic Representation
- Symbolic Backward Refinement

**Preliminaries** 

- Optimisations
- 4 Evaluation
- Conclusion

# Probabilistic Automaton (Example)





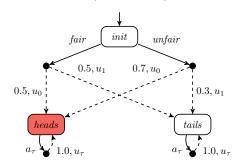
#### Probabilistic Reachability

 $\bullet$  reachability depends on strategy  $\sigma$  of resolving non-determinism

$$Pr_{\mathcal{A}}^{\sigma}(\lozenge G)$$

Motivation

# Probabilistic Automaton (Example)

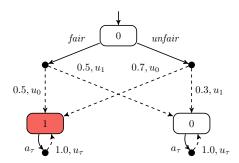


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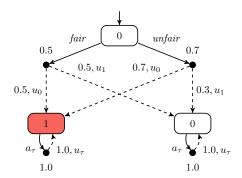
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fixed point characterisation of extremal probabilities

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Motivation



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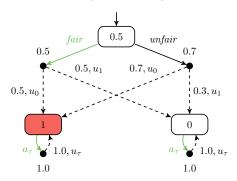
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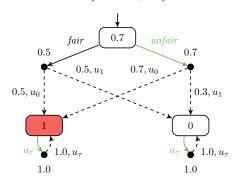


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$$Pr_{\mathcal{A}}^{-}(\lozenge G) \leq Pr_{\mathcal{A}}^{\sigma}(\lozenge G)$$

fixed point characterisation of extremal probabilities



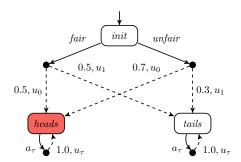
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fixed point characterisation of extremal probabilities

Optimisations

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module simple
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    [a] phase=0
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 $\rightarrow$  (0, -1) Legend: (phase, run)

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Legend: (phase, run)

Optimisations

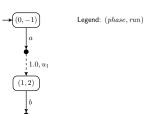
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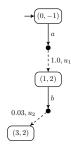
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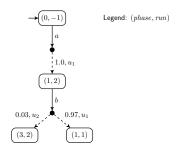
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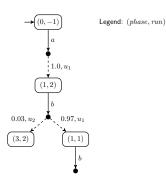
Optimisations

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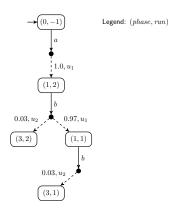
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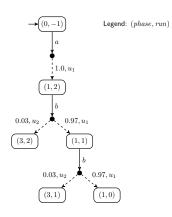
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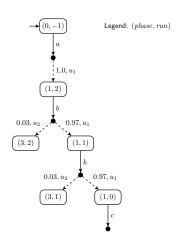
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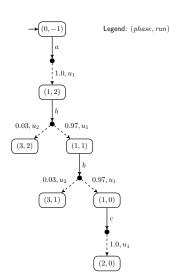


Motivation

Optimisations

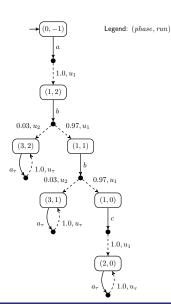
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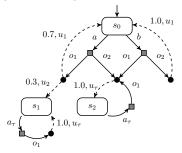
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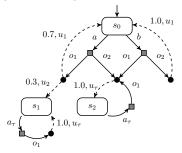




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**Preliminaries** 

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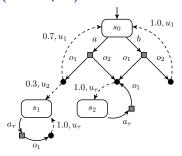


### Probabilistic Reachability

Reachability probability  $Pr_{\mathcal{G}}^{\sigma_{1},\sigma_{2}}\left(\lozenge G\right)$  depends on strategy-pair  $(\sigma_{1},\sigma_{2})$ 

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Outline

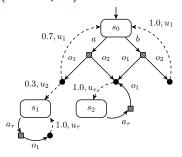


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## Extremal probabilities

Outline



### Probabilistic Reachability

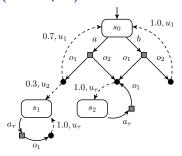
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## Extremal probabilities

$$Pr_{\mathcal{G}}^{-,-}\left(\Diamond G\right)$$

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Outline



### Probabilistic Reachability

Reachability probability  $Pr_{\mathcal{C}}^{\sigma_1,\sigma_2}(\lozenge G)$  depends on strategy-pair  $(\sigma_1,\sigma_2)$ 

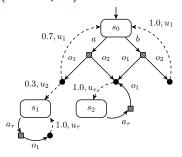
## Extremal probabilities

$$Pr_{\mathcal{G}}^{-,-}\left(\lozenge G\right)$$

$$Pr_{\mathcal{G}}^{-,-}(\lozenge G) \qquad Pr_{\mathcal{G}}^{-,+}(\lozenge G)$$

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Outline



#### Probabilistic Reachability

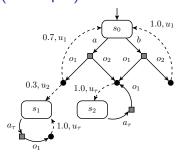
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## Extremal probabilities

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Outline



### Probabilistic Reachability

Reachability probability  $Pr_{\mathcal{G}}^{\sigma_1,\sigma_2}\left(\lozenge G\right)$  depends on strategy-pair  $(\sigma_1,\sigma_2)$ 

## Extremal probabilities

$$Pr_{\mathcal{G}}^{-,-}(\lozenge G) \qquad Pr_{\mathcal{G}}^{-,+}(\lozenge G)$$

$$Pr_{\mathcal{G}}^{+,-}(\lozenge G)$$
  $Pr_{\mathcal{G}}^{+,+}(\lozenge G)$ 

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Motivation

## Observation

Outline

In practice, state spaces exhibit symmetries



Conclusion

## **MTBDD**

Motivation

### Observation

Outline

In practice, state spaces exhibit symmetries

 $\Rightarrow$  exploit by employing symbolical representation

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Preliminaries Symbolic Backward Refinement Optimisations Evaluation

#### **MTBDD**

Motivation

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In practice, state spaces exhibit symmetries

⇒ exploit by employing symbolical representation

# Multi-Terminal Binary Decision Diagram

DAG  $\mathfrak D$  representing a function  $f_{\mathfrak D}:\mathbb B^n o \mathbb R$  with finite range

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Conclusion

Motivation

#### Observation

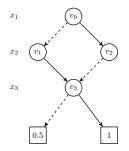
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## Multi-Terminal Binary Decision Diagram

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Motivation

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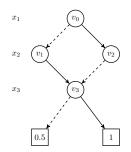
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$x_1$	$x_2$	$x_3$	$f_{\mathfrak{D}}$
0	0	0	0
0	0	1	0
0	1	0	0.5
0	1	1	1
1	0	0	0.5
1	0	1	1
1	1	0	0
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Motivation

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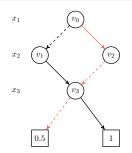
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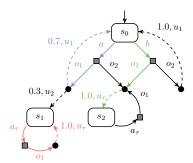
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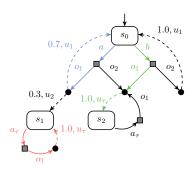
# Stochastic Game as MTBDD (Example)



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Optimisations

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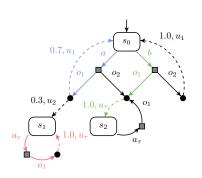
## **Encoding Excerpt**

$$\delta(s_0, b, o_1, u_\tau, s_2) = 1.0$$

Outline

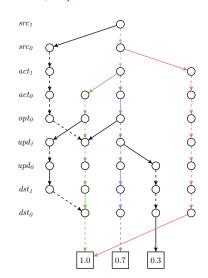
Optimisations

# Stochastic Game as MTBDD (Example)



## **Encoding Excerpt**

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Outline

#### Partition Abstraction

Partition PA's state space S into blocks Q:

$$S = \biguplus_{B \in Q} B$$

# Menu-game: Concept

#### Partition Abstraction

Partition PA's state space S into blocks Q:

$$S = \biguplus_{B \in Q} B$$

### Non-determinism of Model & Abstraction

- merge non-determinism
- distinguish non-determinism

# Menu-game: Concept

#### Partition Abstraction

Partition PA's state space S into blocks Q:

$$S = \biguplus_{B \in Q} B$$

### Non-determinism of Model & Abstraction

- merge non-determinism ⇒ yields PA
- distinguish non-determinism

#### Partition Abstraction

Partition PA's state space S into blocks Q:

$$S = \biguplus_{B \in Q} B$$

### Non-determinism of Model & Abstraction

- ullet merge non-determinism  $\Rightarrow$  yields PA
- distinguish non-determinism ⇒ yields a SG

Outline

## Partition Abstraction

Partition PA's state space S into blocks Q:

$$S = \biguplus_{B \in Q} B$$

#### Non-determinism of Model & Abstraction

- merge non-determinism ⇒ yields PA
- ullet distinguish non-determinism  $\Rightarrow$  yields a SG

# Over-approximation

[Wachter, 2011]

$$Pr_{\mathcal{G}}^{-,-}(\lozenge G^{\#}) \leq Pr_{\mathcal{A}}^{-}(\lozenge G) \leq Pr_{\mathcal{G}}^{-,+}(\lozenge G^{\#})$$

### Partition Abstraction

Partition PA's state space S into blocks Q:

$$S = \biguplus_{B \in Q} B$$

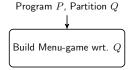
### Non-determinism of Model & Abstraction

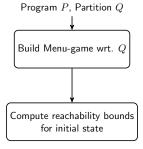
- merge non-determinism ⇒ yields PA
- ullet distinguish non-determinism  $\Rightarrow$  yields a SG

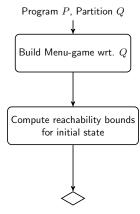
# Over-approximation

[Wachter, 2011]

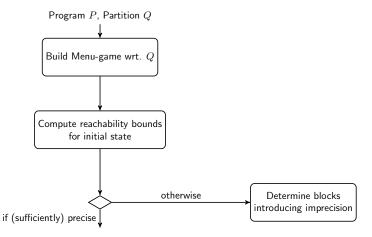
$$Pr_{\mathcal{G}}^{-,-}\left(\lozenge G^{\#}\right) \leq Pr_{\mathcal{A}}^{-}\left(\lozenge G\right) \leq Pr_{\mathcal{G}}^{-,+}\left(\lozenge G^{\#}\right)$$
  
 $Pr_{\mathcal{G}}^{+,-}\left(\lozenge G^{\#}\right) \leq Pr_{\mathcal{A}}^{+}\left(\lozenge G\right) \leq Pr_{\mathcal{G}}^{+,+}\left(\lozenge G^{\#}\right)$ 



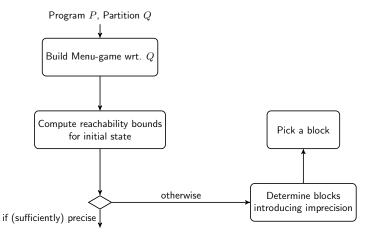


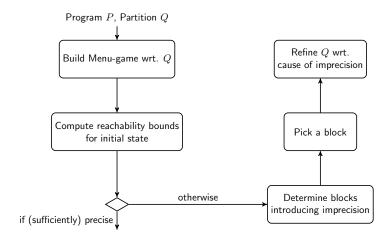


Outline



Outline

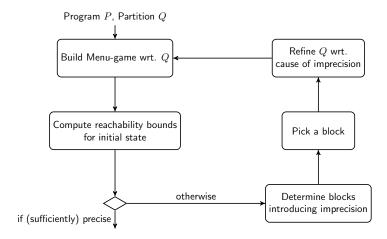


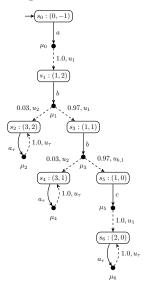


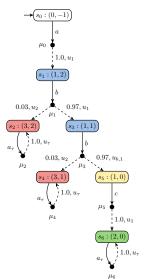
Optimisations

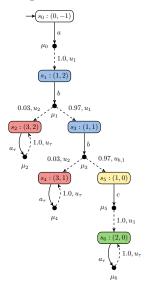
Conclusion

### Backward Refinement Scheme

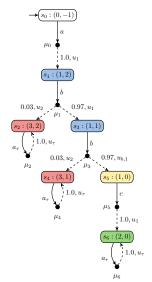






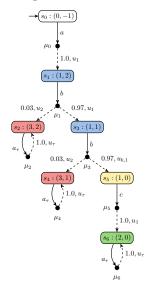




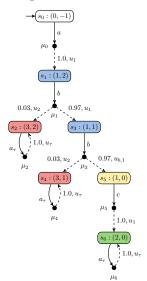


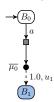


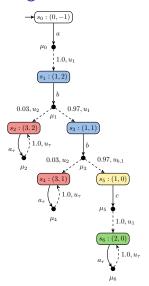
12 / 35

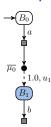


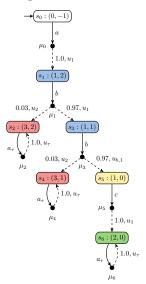


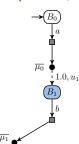


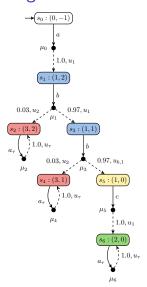


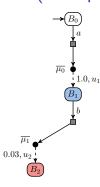


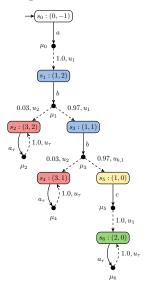


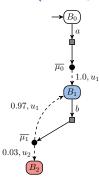


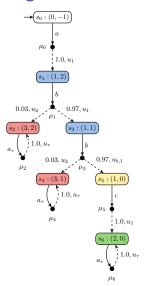


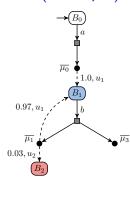


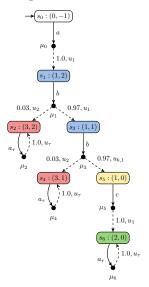


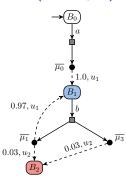


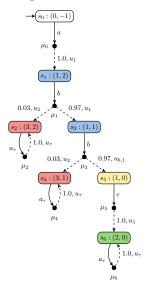


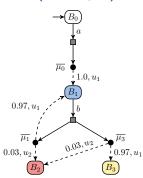






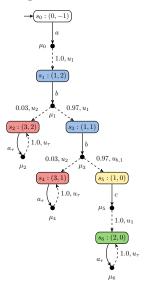


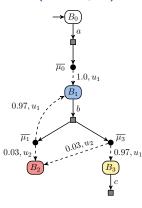


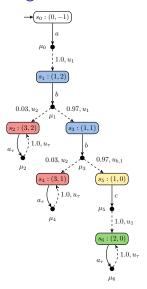


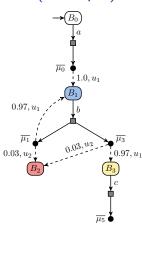
Outline

Conclusion

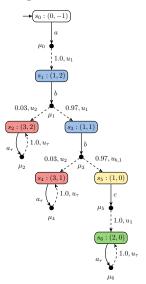


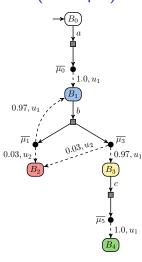






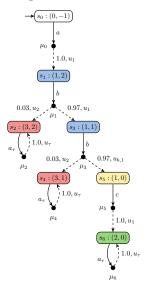
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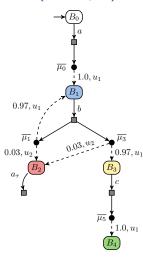


Conclusion

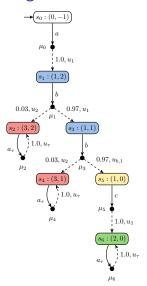
# Menu-game: Construction from PA (Example)

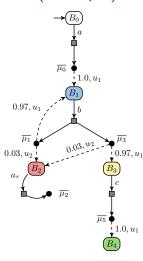


**Preliminaries** 

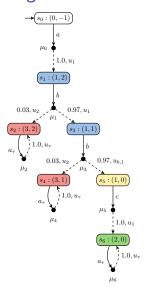


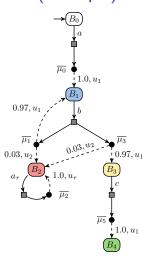
Optimisations



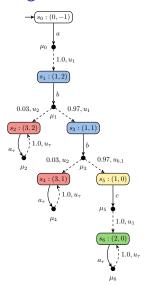


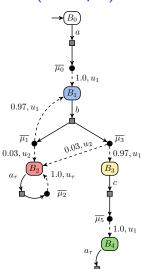
Optimisations

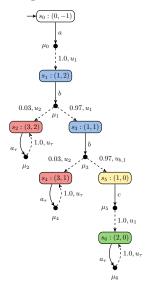


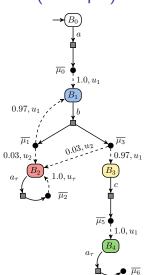


Motivation



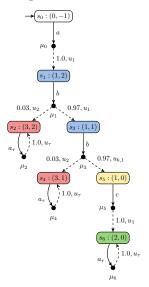


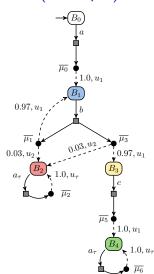




Conclusion

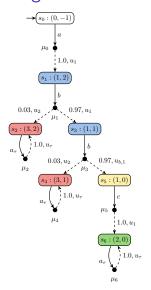
# Menu-game: Construction from PA (Example)





Optimisations

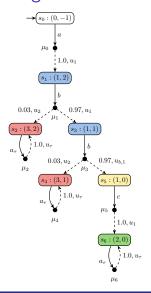
## Menu-game: Predicate Abstraction



### **Predicate**

Boolean expression over a program's variables

## Menu-game: Predicate Abstraction



### Predicate

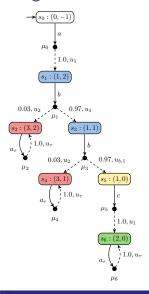
Boolean expression over a program's variables

## Predicates induce partitioning

$$\mathcal{P} = \{phase = 0, phase = 1, phase = 2, phase = 3, run > 0\}$$

Conclusion

## Menu-game: Predicate Abstraction



### Predicate

Boolean expression over a program's variables

### Predicates induce partitioning

$$\mathcal{P} = \{phase = 0, phase = 1, phase = 2, phase = 3, run > 0\}$$

#### induces the partition:

$$\circ phase = 0$$

$$phase = 1, run > 0$$

$$ophase = 1$$

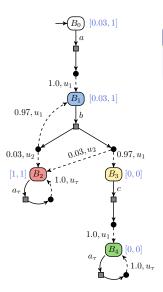
$$phase = 2, run > 0$$

• 
$$phase = 3, run > 0$$

## Refinement

Outline

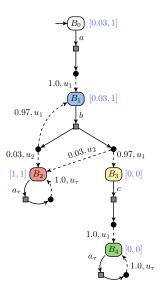
Motivation



## Idea

Split pivot blocks, which introduce imprecision

D. Bohlender



### Idea

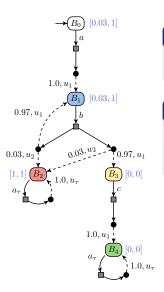
Split *pivot blocks*, which introduce imprecision

### Observations

 deviation alone does not indicate a block being pivot

D. Bohlender

### Refinement

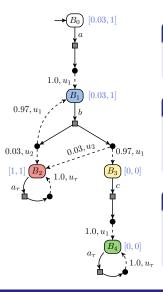


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Split *pivot blocks*, which introduce imprecision

### Observations

- deviation alone does not indicate a block being pivot
- ⇒ differing player 2 strategies do



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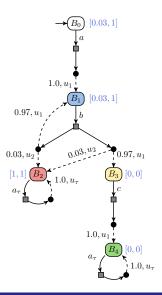
### Observations

- deviation alone does not indicate a block being pivot
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#### Refinement Predicates

 derived from update leading to different blocks

### Refinement



### Idea

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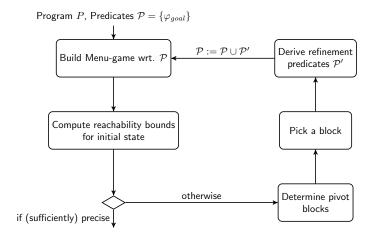
### Observations

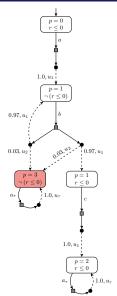
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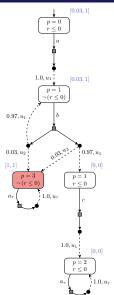
### Refinement Predicates

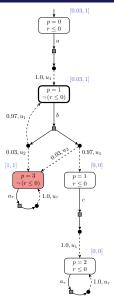
- derived from update leading to different blocks
- ⇒ splitting corresponding behaviours

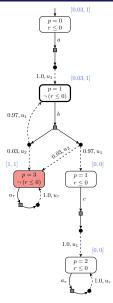
## Reminder

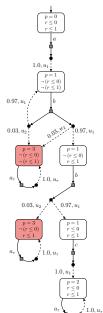


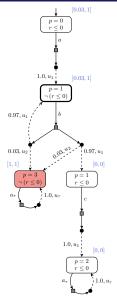


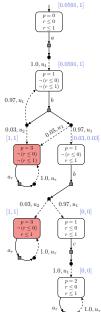


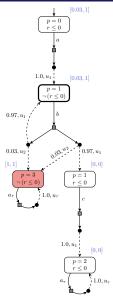


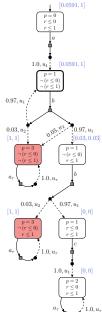


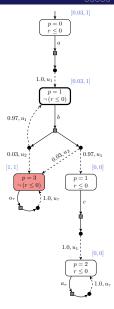


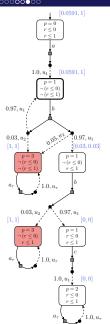


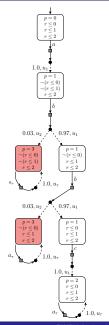


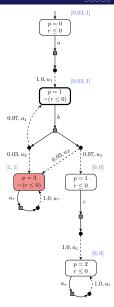


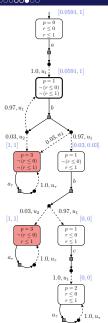


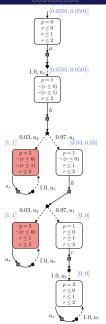




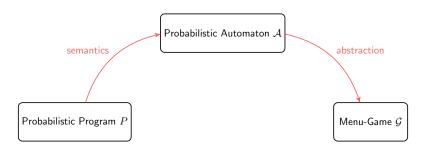






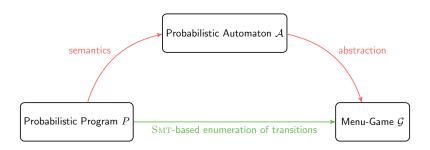


# Motivating SMT-based Construction



D. Bohlender RWTH Aachen University

# Motivating SMT-based Construction



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#### Consider

[a] 
$$x > 0 \rightarrow 0.7$$
:  $(x' = x + 1) + 0.3$ :  $(y' = x)$ 

$$\mathcal{P} = \{x \text{ is odd}, y \text{ is odd}\}$$

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**Preliminaries** 

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$$\land (b_0^{src} \Leftrightarrow x \text{ is odd}) \land (b_1^{src} \Leftrightarrow y \text{ is odd})$$

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$$\wedge \; \left(b_0^{dst_1} \Leftrightarrow x+1 \text{ is odd}\right) \wedge \left(b_1^{dst_1} \Leftrightarrow y \text{ is odd}\right)$$

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#### Abstract Transition Constraint

$$\land (b_0^{src} \Leftrightarrow x \text{ is odd}) \land (b_1^{src} \Leftrightarrow y \text{ is odd})$$

$$\wedge (b_0^{dst_1} \Leftrightarrow x+1 \text{ is odd}) \wedge (b_1^{dst_1} \Leftrightarrow y \text{ is odd})$$

$$\land (b_0^{dst_2} \Leftrightarrow x \text{ is odd}) \land (b_1^{dst_2} \Leftrightarrow x \text{ is odd})$$

### Interpretation

- $\bullet$   $(b_0^{src}, b_1^{src}) = (1, 0)$
- $\bullet$   $(b_0^{dst_1}, b_1^{dst_1}) = (0, 0)$
- $\bullet$   $(b_0^{dst_2}, b_1^{dst_2}) = (1, 1)$

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#### Abstract Transition Constraint

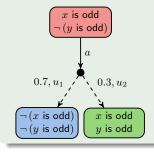
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Outline

### Observation

Motivation



#### Observation

Commands are often only related to a subset of all predicates

 $\mathcal{P}_{u_j}^{src}$  indicate the (in)validity of predicates in the successor  $B_j$ 

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#### Observation

Motivation

- $\mathcal{P}_{u_j}^{src}$  indicate the (in)validity of predicates in the successor  $B_j$ , e.g. share variable with assignment
- $\mathcal{P}_{u_j}^{dst}$  whose validity in successor blocks may be affected by  $u_j$

Optimisations

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## Relevant Predicates

Outline

#### Observation

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- $\Rightarrow$  irrelevant destination predicates retain their value

Motivation

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- ⇒ irrelevant destination predicates retain their value

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$$x > 0 \to 0.7 : (x' = x + 1) + 0.3 : (y' = x)$$

$$x > 0 \land (b_0^{src} \Leftrightarrow x \text{ is odd}) \land (b_1^{src} \Leftrightarrow y \text{ is odd})$$

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Motivation

Commands are often only related to a subset of all predicates

- $\mathcal{P}_{u_j}^{src}$  indicate the (in)validity of predicates in the successor  $B_j$ , e.g. share variable with assignment
- $\mathcal{P}_{u_j}^{dst}$  whose validity in successor blocks may be affected by  $u_j$ , e.g. contain assignment variable.
- ⇒ irrelevant destination predicates retain their value

$$\begin{split} [a] \ x > 0 &\to 0.7 : (x' = x + 1) + 0.3 : (y' = x) \\ x > 0 \ \land \ (b_0^{src} \Leftrightarrow x \text{ is odd}) \land (b_1^{src} \Leftrightarrow y \text{ is odd}) \\ &\land \ (b_0^{dst_1} \Leftrightarrow x + 1 \text{ is odd}) \land (b_1^{dst_1} \Leftrightarrow y \text{ is odd}) \\ &\land \ (b_0^{dst_2} \Leftrightarrow x \text{ is odd}) \land (b_1^{dst_2} \Leftrightarrow x \text{ is odd}) \end{split}$$

#### Observation

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$$\begin{split} [a] \ x > 0 &\rightarrow 0.7 : (x' = x + 1) + 0.3 : (y' = x) \\ x > 0 \wedge \left(b_0^{src} \Leftrightarrow x \text{ is odd}\right) \wedge \left(b_1^{src} \Leftrightarrow y \text{ is odd}\right) \\ & \wedge \left(b_0^{dst_1} \Leftrightarrow x + 1 \text{ is odd}\right) \wedge \left(b_1^{dst_1} \Leftrightarrow y \text{ is odd}\right) \\ & \wedge \left(b_0^{dst_2} \Leftrightarrow x \text{ is odd}\right) \wedge \left(b_1^{dst_2} \Leftrightarrow x \text{ is odd}\right) \end{split}$$

Outline

#### Observation

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- $\Rightarrow$  irrelevant destination predicates retain their value

# Simplify Transition Constraint

$$\begin{split} [a] \ x > 0 \to 0.7 : (x' = x + 1) + 0.3 : (y' = x) \\ x > 0 \wedge (b_0^{src} \Leftrightarrow x \text{ is odd}) \wedge (\underline{b_1^{src}} \Leftrightarrow y \text{ is odd}) \\ \wedge (b_0^{dst_1} \Leftrightarrow x + 1 \text{ is odd}) \wedge (\underline{b_1^{dst_1}} \Leftrightarrow y \text{ is odd}) \\ \wedge (\underline{b_0^{dst_2}} \Leftrightarrow x \text{ is odd}) \wedge (b_1^{dst_2} \Leftrightarrow x \text{ is odd}) \\ \Rightarrow \text{ extend solutions with } b_1^{src} \Leftrightarrow b_1^{dst_1} \text{ and } b_0^{src} \Leftrightarrow b_0^{dst_2} \end{split}$$

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## Observation

Outline

Motivation

Refinement can only split blocks but never introduce "new" ones

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### Reachable Blocks as Constraint

## Observation

Motivation

Refinement can only split blocks but never introduce "new" ones

⇒ new transition constraint solutions extend the old ones



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Motivation

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#### New Solutions are Extensions

Let the old solution have only three source blocks:

$$(b_0^{src}, b_1^{src}, b_2^{src}) \in \{(0, 0, 1), (0, 1, 1), (1, 0, 0)\}$$

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**Preliminaries** 

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#### Idea

Extend transition constraint with reachable blocks constraints

**Preliminaries** 

Variables' Ranges

⇒ extend constraint with variables' domains

**Exploit** Incrementality

**Predicate** Decomposition

Variables' Ranges

Motivation

- ⇒ extend constraint with variables' domains
- $\bullet$  e.g. for  $x \in \{0,1,2\}$  add  $0 \le x \land x \le 2$

Exploit Incrementality

Predicate Decomposition

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- **Exploit** incremental checking faster than starting from scratch

Predicate Decomposition

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Optimisations

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transition constraint grows monotonously

**Predicate** Decomposition

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⇒ split spuriously coupled variables

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# Unrelated Commands

new predicate often not relevant for all commands

## Variables' Ranges

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- ⇒ reuse previous solutions

# Pre-computation

Outline

Motivation

## Observations

ullet value iteration may not yield reachability probability to be exactly 1

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**Evaluation** 

Conclusion

# Pre-computation

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- convergence to 0 or 1 may be slow

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### Idea

Motivation

Extend pre-computation algorithms for PA to Menu-games



Optimisations

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Conclusion

#### Pre-computation

#### Observations

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#### Idea

Extend pre-computation algorithms for PA to Menu-games

" $\sigma_1$ "	" $\sigma_2$ "	Prob0	Prob1
_	_	EE	AA
_	+		
+	_		
+	+		

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**Preliminaries** 

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_	+	$\mathrm{EA}$	AE
+	_	AE	EA
+	+	AA	EE

#### Observations

- ullet value iteration may not yield reachability probability to be exactly 1
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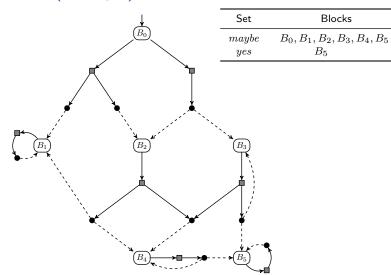
#### Idea

Motivation

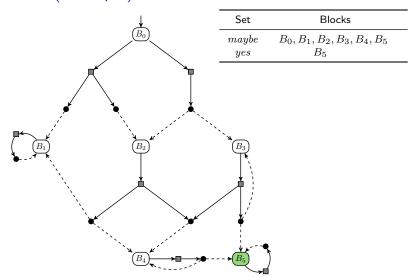
Extend pre-computation algorithms for PA to Menu-games

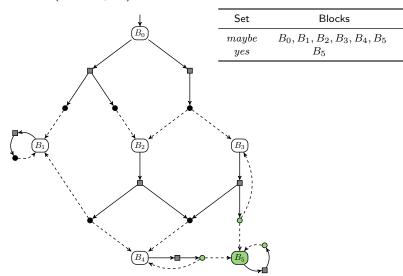
" $\sigma_1$ "	" $\sigma_2$ "	Prob0	Prob1
_	_	EE	AA
_	+	$\mathrm{EA}$	AE
+	_	AE	EA
+	+	AA	EE

#### PROB1EA (Example)

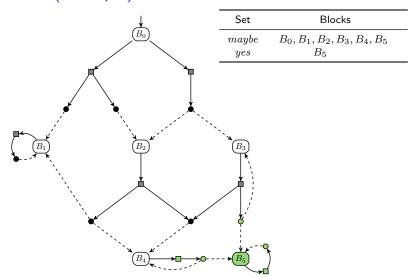


**Preliminaries** 

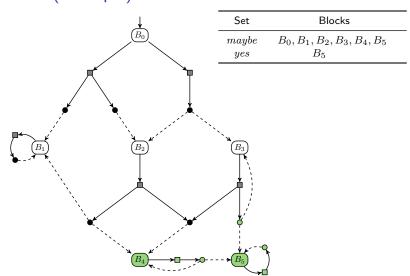




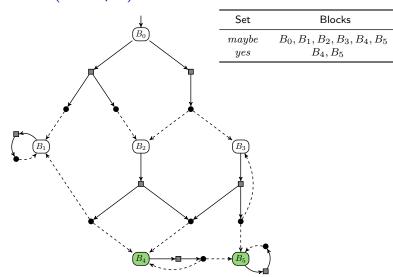
## PROB1EA (Example)



# PROB1EA (Example)



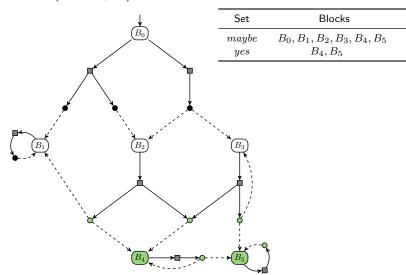
**Preliminaries** 



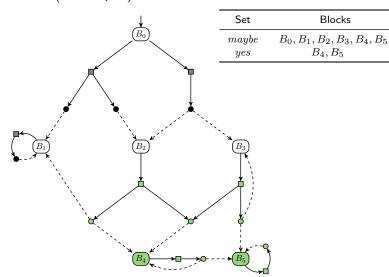
**Evaluation** 

# PROB1EA (Example)

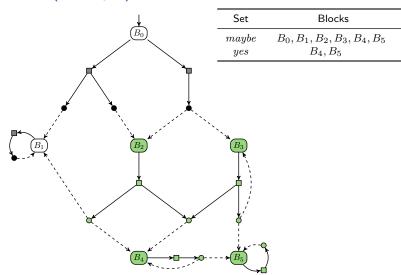
**Preliminaries** 

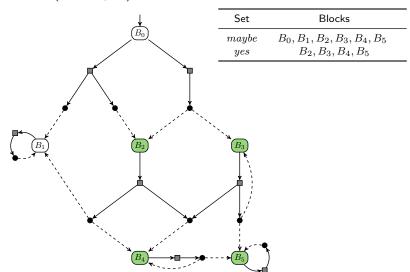


# PROB1EA (Example)



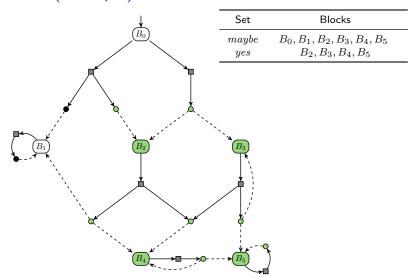
# PROB1EA (Example)



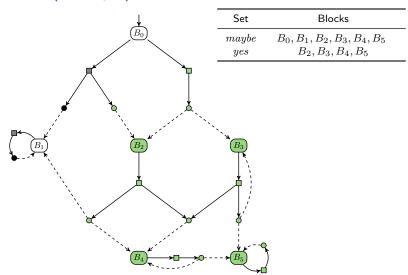


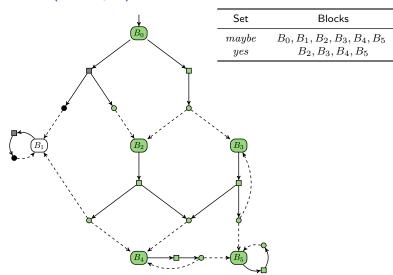
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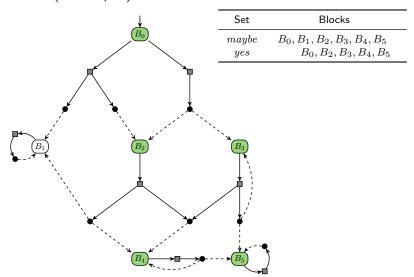
#### PROB1EA (Example)





Conclusion

### PROB1EA (Example)

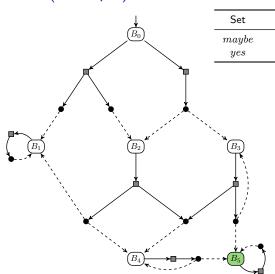


**Blocks** 

 $B_0, B_2, B_3, B_4, B_5$ 

 $B_5$ 

## PROB1EA (Example)

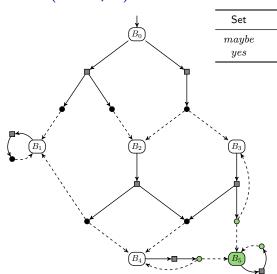


**Blocks** 

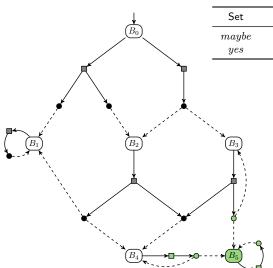
 $B_0, B_2, B_3, B_4, B_5$ 

 $B_5$ 

## PROB1EA (Example)



## PROB1EA (Example)

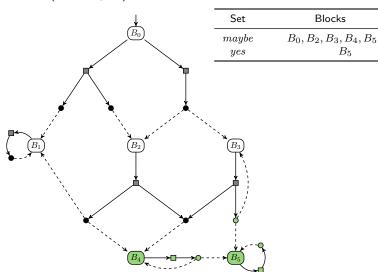


 $\begin{array}{ccc} \text{Set} & \text{Blocks} \\ \hline maybe & B_0, B_2, B_3, B_4, B_5 \\ yes & B_5 \end{array}$ 

**Blocks** 

 $B_5$ 

## PROB1EA (Example)

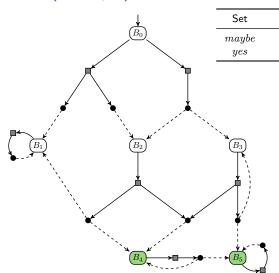


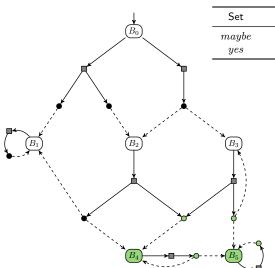
**Blocks** 

 $B_0, B_2, B_3, B_4, B_5$ 

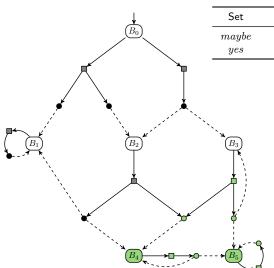
 $B_4, B_5$ 

### PROB1EA (Example)

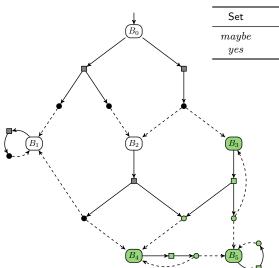




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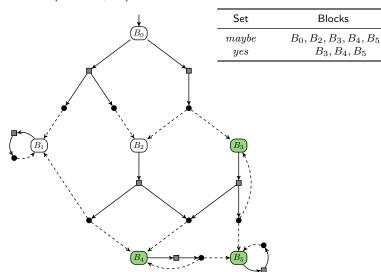


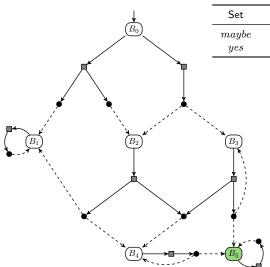
 $\begin{array}{ccc} \text{Set} & \text{Blocks} \\ \hline \textit{maybe} & B_0, B_2, B_3, B_4, B_5 \\ \textit{yes} & B_4, B_5 \end{array}$ 



 $\begin{array}{ccc} \text{Set} & \text{Blocks} \\ maybe & B_0, B_2, B_3, B_4, B_5 \\ yes & B_4, B_5 \end{array}$ 

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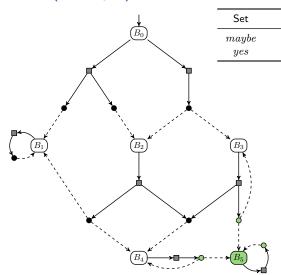


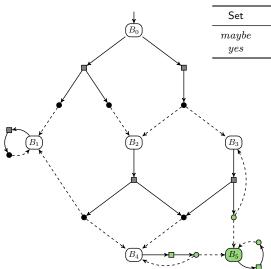
Set	Blocks	
maybe	$B_3, B_4, B_5$	
ues	$B_{5}$	

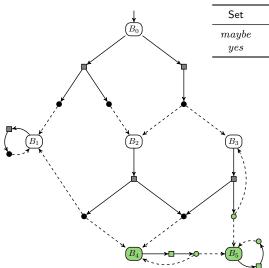
**Blocks** 

 $B_3, B_4, B_5$  $B_5$ 

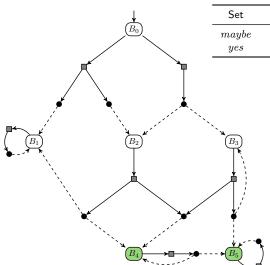
## PROB1EA (Example)







Set	Blocks	
maybe	$B_3, B_4, B_5$	
yes	$B_5$	



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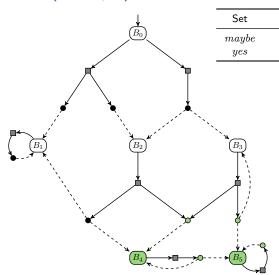
Evaluation

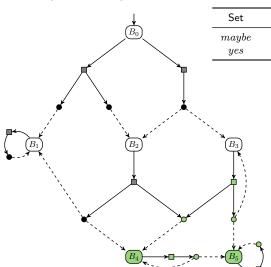
**Blocks** 

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## PROB1EA (Example)

**Preliminaries** 





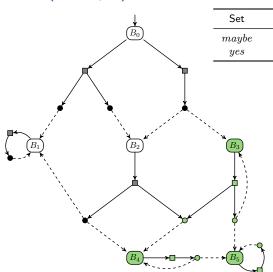
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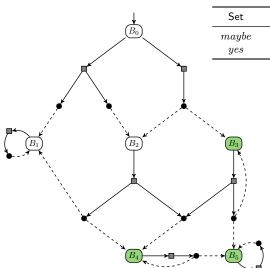
## PROB1EA (Example)

**Preliminaries** 



Motivation

# PROB1EA (Example)



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### Other tweaks

Motivation

Reuse reachability

avoid starting value iteration from scratch

Remove goal transitions

### Other tweaks

# Reuse reachability

- avoid starting value iteration from scratch
- ⇒ reuse previous refinement iteration results (where applicable)

Remove goal transitions

### Other tweaks

Outline

Motivation

# Reuse reachability

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# Remove goal transitions

focus on probabilistic reachability

Outline

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#### Remove goal transitions

- focus on probabilistic reachability
- irrelevant what happens once goal is reached

# Reuse reachability

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- ⇒ reuse previous refinement iteration results (where applicable)

# Remove goal transitions

- focus on probabilistic reachability
- irrelevant what happens once goal is reached
- ⇒ remove outgoing transitions of goal blocks

Outline

Motivation

### Prototypical Implementation

ullet uses  $\operatorname{Storm}$ 's parser, expressions and can use explicit value iteration

## Prototypical Implementation

- $\bullet$  uses  $\operatorname{Storm}$  's parser, expressions and can use explicit value iteration
- $\bullet$  uses Z3 as  $SmT\mbox{-solver}$  and CUDD as MTBDD-library

Motivation

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Motivation

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Outline

- corner cases of backward refinement not documented
- vague (to not existent) description of PASS's implementation details
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Motivation

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### Case Studies

• 4 case studies (focus on two here)

Motivation

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### Case Studies

- 4 case studies (focus on two here)
- measured impact of optimisations on run time and game sizes

Outline

Motivation

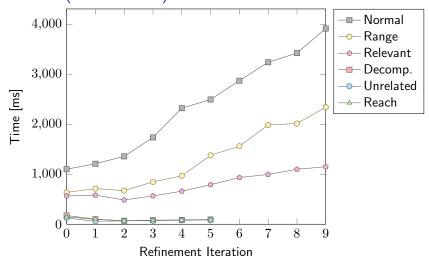
### Prototypical Implementation

- $\bullet$  uses  $\operatorname{Storm}$  's parser, expressions and can use explicit value iteration
- $\bullet$  uses Z3 as Smt-solver and CUDD as MTBDD-library
- obstacles:
  - corner cases of backward refinement not documented
  - vague (to not existent) description of PASS's implementation details
  - standard MTBDD operations not sufficient for strategy computation
  - . . .
- final implementation has  $\approx 6000$  lines of code (18.000 committed)

### Case Studies

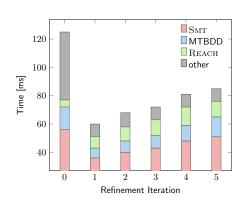
- 4 case studies (focus on two here)
- measured impact of optimisations on run time and game sizes
- evaluated symbolic vs. explicit analysis (memory usage & run time)

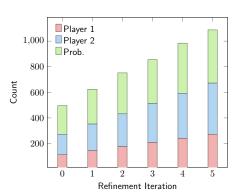
# Consensus (Abstraction)



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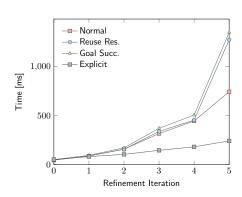
Outline

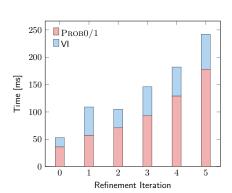




Motivation

# Consensus (Analysis)



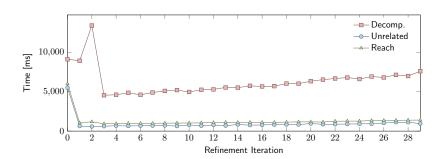


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# WLAN (Abstraction)

Outline

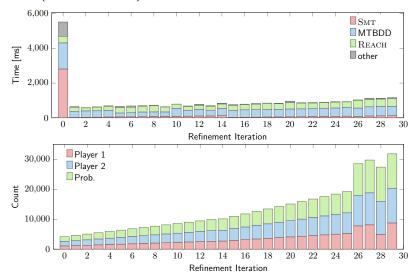
Motivation



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# WLAN (Abstraction)

Outline



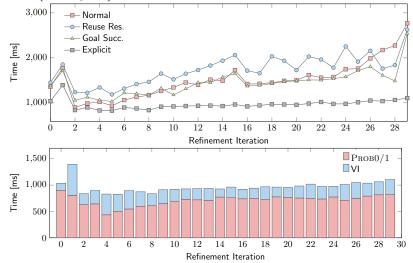
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Motivation

# WLAN (Analysis)

Outline

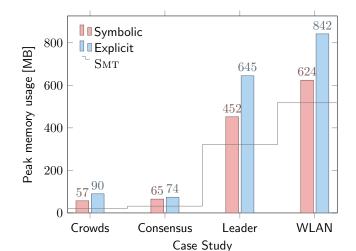
Motivation



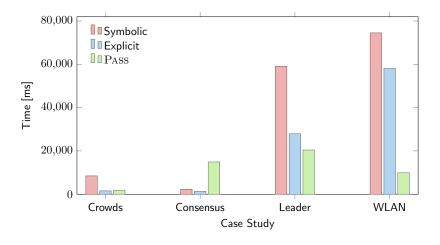
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**Evaluation** 

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## Symbolic vs. Explicit vs. PASS



Motivation

### Summary

• Menu-game as over-approximation of a PA



### Summary

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- several optimisations for both abstraction and analysis

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Optimisations

Conclusion

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#### Future work

topological symbolic value iteration

34 / 35

Optimisations

## Conclusion

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- exploit modularity of probabilistic programs

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Motivation

Outline

Thanks for your attention!

Interested in details? Suggestions?

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