# Quantifying Information Leakage of Probabilistic Programs Using the PRISM Model Checker

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#### Confidentiality





#### Common mechanisms for confidentiality:

Cryptography Access control Firewall





Introduction

#### Information leakage

secret variables

public variables









l := h |  $(1100)_b$ 

2 rightmost bits of h are leaked into 1





## Introduction

#### Information leakage

1 bit of h is leaked into 11





1. An automated method:

- Modeling programs by Markov chains,
- Computing joint probabilities of the program's secrets and public outputs,
- Calculating the exact value of information leakage.





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## Contributions

#### 2. PRISM-Leak

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#### A tool for evaluating secure information flow of concurrent probabilistic programs

leakage	prism	information-leakage	binary-decision-diagrams	prism-language	security	security-tool	concurrent-probabilistic-programs
confidenti	ality						
G	) 32 comr	nits	ဖို <b>2</b> branches	🛇 2 releases		🎎 1 contr	ributor 👌 GPL-3





Contributions

3. Case study:

the grades protocol



## Contents





Markov Chain  $\mathcal{M} = (S, \mathbf{P}, \zeta)$ 





Markov Chain  $\mathcal{M} = (S, \mathbf{P}, \zeta)$ 





Markov Chain  $\mathcal{M} = (S, \underline{\mathbf{P}}, \zeta)$ 







Quantifying Information Leakage of Probabilistic Programs ...



h = 1

#### Occurrence probability of a path







#### $\mathcal{L}(\mathcal{M}) = \text{initial uncertainty } - \text{remaining uncertainty}$







$$\mathcal{H}(\mathcal{X}) = -\Sigma_{x \in \mathcal{X}} \Pr(\mathcal{X} = x) \log_2 \Pr(\mathcal{X} = x)$$





#### $\mathcal{L}(\mathcal{M}) = \text{initial uncertainty } - \text{remaining uncertainty}$

$$\mathcal{L}(\mathcal{M}) = \mathcal{H}(h) - \mathcal{H}(h \mid o)$$





#### Initial uncertainty

$$\mathcal{H}(h) = -\sum_{\overline{h} \in h} Pr\left(h = \overline{h}\right) \cdot \log_2 Pr\left(h = \overline{h}\right)$$





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### The proposed method

Remaining uncertainty

$$\mathcal{H}(h \mid o) = -\sum_{\overline{o} \in o} Pr(o = \overline{o}) \cdot \mathcal{H}(h \mid o = \overline{o})$$
$$-\sum_{\overline{h} \in h} Pr(h = \overline{h} \mid o = \overline{o}) \cdot \log_2 Pr(h = \overline{h} \mid o = \overline{o})$$
$$\sum_{\overline{h} \in h} Pr(h = \overline{h}, o = \overline{o}) \qquad \qquad \frac{\Pr(h = \overline{h}, o = \overline{o})}{\Pr(o = \overline{o})}$$



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## The proposed method

Remaining uncertainty

$$\mathcal{H}(h \mid o) = -\sum_{\overline{o} \in o} Pr(o = \overline{o}) \cdot \mathcal{H}(h \mid o = \overline{o})$$
$$-\sum_{\overline{h} \in h} Pr(h = \overline{h} \mid o = \overline{o}) \cdot \log_2 Pr(h = \overline{h} \mid o = \overline{o})$$
$$\sum_{\overline{h} \in h} Pr(h = \overline{h}, o = \overline{o})$$
$$\underbrace{\Pr(h = \overline{h}, o = \overline{o})}_{\Pr(o = \overline{o})}$$



$$\sum_{\overline{h}\in h} Pr\left(h = \overline{h}, o = \overline{o}\right) =$$

$$\sum_{s_0 \in Init(\mathcal{M}), \ s_n = \langle \overline{o}, \overline{h}, ... \rangle} Pr(\pi = s_0 \dots s_n)$$





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Input: finite MC  $\mathcal{M}$ 

*Output*: a map containing the joint probabilities Pr(h, o)

- 1: Let ohMap be an empty higher-order map function from  $\overline{o}$  to  $\overline{h}$  to  $Pr(h = \overline{h}, o = \overline{o})$ ; *// i.e.*  $ohMap : \overline{o} \mapsto (\overline{h} \mapsto Pr(h = \overline{h}, o = \overline{o}))$
- 2: Let  $\pi$  be an empty list of states for storing a path;
- 3: for  $s_0$  in  $Init(\mathcal{M})$  do
- 4: EXPLOREPATHS( $s_0, \pi, ohMap$ );
- 5: return ohMap;



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# The proposed method

6:	function EXPLOREPATHS(s, $\pi$ , $ohMap$ )
	<i>II add state s to the current path from the initial state</i>
7:	$\pi$ .add(s);
	// found a path stored in $\pi$
8:	if s is a terminating state then
9:	// assume $s = \langle \overline{o}, \overline{h}, \cdot, \cdot \rangle$
	// define $hMap$ as $Pr(h, o = \overline{o})$
10:	if $\overline{o}$ not in $ohMap$ then
11:	Let $hMap$ be an empty map from
	$\overline{h}$ to $Pr(h = \overline{h}, o = \overline{o});$
12:	else
13:	$hMap = ohMap.get(\overline{o});$
14:	if $\overline{h}$ not in $hMap$ then
15:	$prob = Pr(\pi);$
16:	else
17:	$prob = Pr(\pi) + hMap.get(\overline{h});$
18:	$hMap.put(\overline{h}, prob);$ // Update $hMap$
19:	$ohMap.put(\overline{o}, hMap); // Update ohMap$
20:	else
21:	for $s'$ in $Post(s)$ do
22:	EXPLORE PATHS( $s', \pi, ohMap$ );
	// done exploring from s, so remove it from $\pi$
23:	$\pi$ .pop();
24:	return ;



#### Time complexity:

 $O(2^{n})$ 



## Contents







## Implementation

**PRISM-Leak:** 







The grades protocol

- k students  $s_1, \ldots, s_k$
- secret grades  $g_1, \ldots, g_k$  where  $0 \le g_i < m$
- Goal: computing sum of the grades, without revealing the secret grades to other students





# Case study

#### The grades protocol

- k students  $s_1, \ldots, s_k$
- secret grades  $g_1, \ldots, g_k$  where  $0 \le g_i < m$
- $n = (m-1) \times k + 1$
- $r_i \in [0, n]$
- $d_i = g_i + r_i r_{(i+1)\% k}$
- sum =  $(\sum_i d_i) \% n$







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Case study

#### The grades protocol

	k		The grades prot	locol	The sum of the grades			
m		$\mathcal{M}_{grades}$		Leakage	л Л	Leakage		
		# states	# transitions	(bits)	# states	# transitions	(bits)	
	2	196	228	1.5 (75%)	16	20	1.5	
2	3	3752	4256	1.81 (60.4%)	64	104	1.81	
	4	92496	102480	2.03 (50.8%)	256	528	2.03	
	2	1179	1395	2.2 (69.3%)	36	45	2.2	
3	3	66366	75600	2.53 (53.1%)	216	351	2.53	
	4	439668	597780	2.75 (43.3%)	1296	2673	2.75	
	2	4048	4816	2.66 (66.4%)	64	80	2.66	
4	3	455104	519040	2.98 (49.7%)	512	832	2.98	
	4	3271680	6589440	3.2 (40%)	4096	8448	3.2	

## Contents

Introduction
The proposed method
Implementation and case study
Related work
Conclusion





Chothia et al., 2013

- Tool LeakWatch
- Java programs
- Estimation of the leakage
- Intermediate leakages





#### Klebanov, 2014

- Symbolic execution and self-composition
- Deterministic programs
- Non-automated method





Biondi et al., 2017

- Tool HyLeak
- Sequential programs
- Estimation of the leakage

• No intermediate leakage





Salehi et al., 2019

- Evolutionary algorithm
- Channel capacity
- Concurrent probabilistic programs



## Contents

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## Conclusion

Proposed approach:







- 1. Comparing scalability
- 2. Estimating leakage by statistical methods
- 3. Analyzing case studies in other application domains





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#### Thanks for you attention!

