

Computing Information Flow Using Symbolic Model-Checking

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Outline

Introduction

Preliminaries

Summary Calculation

Computing Information Leakage: Symbolic Algorithms

Moped-QLeak

Demo

Conclusions and Future Work

Thank You

Introduction

- ▶ Quantifying information leakage - Inferring information about inputs by observing public outputs

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- ▶ No leakage \implies Outputs independent of inputs
- ▶ Full leakage \implies Unique input corresponding to given output
- ▶ Comparing leakage across programs - less leakage is desirable

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2. Shannon entropy leakage measures expected number of guesses required to correctly guess the secret input

$$\text{SE}_U(P) = \log |\mathcal{S}| - \frac{1}{|\mathcal{S}|} \sum_{o \in \mathcal{O}} |P^{-1}(o)| \log |P^{-1}(o)|$$

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- ▶ Determine if the NSA paid or not w/o revealing information about cryptographers

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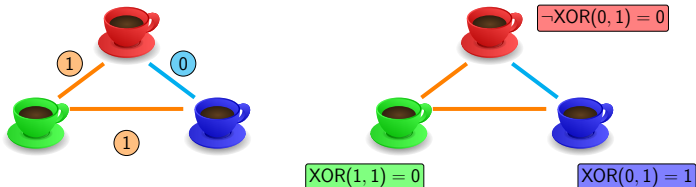
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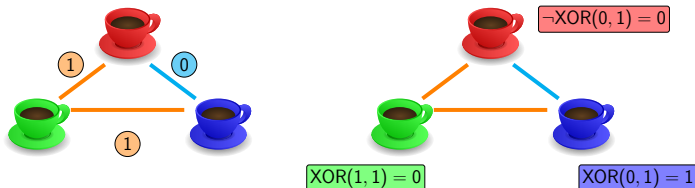
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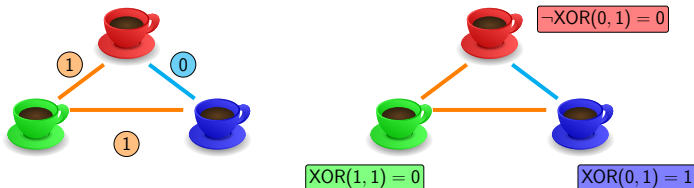


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$\text{XOR}(\text{Announcement}_A, \text{Announcement}_B, \text{Announcement}_C) = 0$
iff
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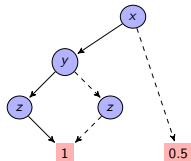
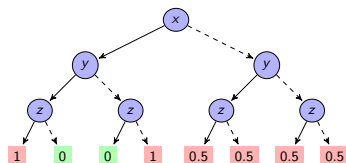
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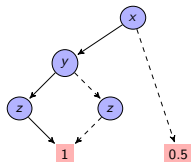
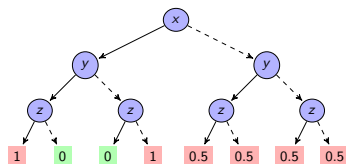
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ADD (up) and its reduced form (bottom)

Computing Summaries: Fixed Point Iteration

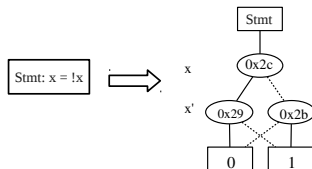
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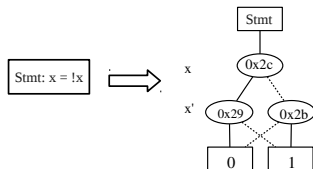
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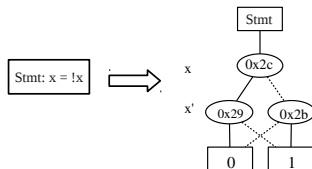
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- ▶ Compose statements
- ▶ Arrive at a fixed point (Summary μ)

Min Entropy : Symbolic Algorithm

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Algorithm 6: Symbolic computation of min-entropy leakage of a probabilistic program

Input: $\mathcal{G}, \mathcal{G}'$ and T_P the summary of P .

Output: $\text{ME}_U(P)$

```
1 begin
2    $T_{\text{out},P} \leftarrow \text{abstract}(\text{max}, \mathcal{G}, T_P)$ 
3    $\text{sum}_{\text{out}} \leftarrow \text{val}(\text{abstract}(+, \mathcal{G}', T_{\text{out},P}))$ 
4    $T_{\text{term},P} \leftarrow \text{abstract}(+, \mathcal{G}', T_P)$ 
5    $\text{sum}_{\text{out}} \leftarrow \text{sum}_{\text{out}} + (1 - \text{val}(\text{abstract}(\text{min}, \mathcal{G}, T_{\text{term},P})))$ ;
6   return  $\log \text{sum}_{\text{out}}$ 
```

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Algorithm 8: Symbolic computation of Shannon entropy leakage of a probabilistic program

Input: $\mathcal{G}, \mathcal{G}'$ and T_P the summary of P .

Output: $\text{SE}_U(P)$

```
1 Let  $n$  be the number of variables in  $\mathcal{G}$ .
2 begin
3    $T_{\text{norm-eq-size},P} \leftarrow \text{divide}(\text{abstract}(+, \mathcal{G}, T_P), 2^n)$ 
4    $\text{val}_{\text{out}} \leftarrow (- \text{val}(\text{abstract}(\star, \mathcal{G}', T_{\text{norm-eq-size},P})))$ 
5    $T_{\text{term},P} \leftarrow \text{abstract}(+, \mathcal{G}', T_P)$ 
6    $\text{prob}_{\text{out,non-term}} \leftarrow (1 - \frac{\text{val}(\text{abstract}(+, \mathcal{G}, T_{\text{term},P}))}{2^n})$ 
7    $\text{val}_{\text{out,non-term}} \leftarrow (- \text{prob}_{\text{out,non-term}} \log \text{prob}_{\text{out,non-term}})$ 
8    $T_{\text{norm-}\star\text{out},P} \leftarrow \text{divide}(\text{abstract}(\star, \mathcal{G}', T_P), 2^n)$ 
9    $\text{val}_{\text{cond}} \leftarrow (-\text{val}(\text{abstract}(+, \mathcal{G}, T_{\star\text{out},P})))$ 
10   $T_{\text{non-term},P} \leftarrow \text{subtract}(1, T_{\text{term},P})$ 
11   $\text{val}_{\text{cond,non-term}} \leftarrow (- \frac{\text{val}(\text{abstract}(\star, \mathcal{G}, T_{\text{non-term-prob},P}))}{2^n})$ 
12  return  $(\text{val}_{\text{out}} + \text{val}_{\text{out,non-term}} - \text{val}_{\text{cond}} - \text{val}_{\text{cond,non-term}})$ 
```

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```
define N 32
define DEFAULT_INT_BITS N
unsigned int var1;
bool g;

module void f(unsigned int v, bool z){
    bool k;
    pchoice
    :: 0.2 -> label2: k = g && z;
    :: 0.8 -> var1 = var1 + v;
    choicep
}

module void main(){
    var1 = 53;
    pchoice
    :: 0.3 -> label1: g = true;
    :: 0.7 -> f(var1, !g);
    choicep
}
```

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- ▶ Time taken in milliseconds
- ▶ Consistently outperforms **sqifc** (Malacaria et. al)

Example	Order	ME	SE	Time	Data types
Illustrative Example	I	3	2.03966e-05	0.215	bool
Electronic Purse	D	2	2	0.009	5 bit integers (Restricted)
Mix and Duplicate	S	16	16	0.041	bool
Binary Search	I	16	16	9.307	bool
Sanity Check	I	4	1.168e-7	0.060	bool
Implicit Flow	D	2.8074	1.757e-07	0.016	30 bit integers
Implicit Flow	D	2.8074	0.003	0.010	15 bit integers
Implicit Flow	D	2.8074	4.67189e-08	0.190	bool
Masked Copy	I	16	16	0.038	bool
Sum Query	D	4.80735	4.35132	0.034	5 bit integers (Restricted)

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Tool demonstration

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