## Synthesis of Adaptive Side-Channel Attacks

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## Overview



## Motivating Example

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High security input (secret): h
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    return 0;
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Always 0 . No information.

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$t=1 \Rightarrow h \leq 1$

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Always 0 . No information.

Side channel:
$t=1 \Rightarrow h \leq I$
$t=2 \Rightarrow h>l$

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t=1 \Rightarrow h \leq 1 \\
t=2 \Rightarrow & \Rightarrow>1
\end{array}
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| 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 |
| :--- | :--- | :--- | :--- | ---: | :--- | :--- | :--- |

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| 1 | $1=6$ |  |  |  |  |  |  |
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Too few divisions.

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Unbalanced divisions.

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## Find the Best Attack!

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 How?
## Our Approach

$7 / 29$

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- Results in symbolic tree (attack tree).


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$h$ and all l-choices symbolic constraints between $h$ and / symbolic

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# Finding Best Attack Tree Method 1 

## Maximizing Number of Partition Divisions

```
foo(int l,int h)
    if (l<0)
        if (h<0) sleep(1)
        else if (h<5) sleep(2)
        else sleep(3)
    else
    if (h>1) sleep(4)
    else sleep(5)
```

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- Find an assignment for I and $h_{i}$ that maximizes the number of satisfiable constraints.
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MAX-SMT Problem: Find an assignment of values to variables that maximizes the number of simultaneously satisfied clauses.

# Finding Best Attack Tree Method 2 

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Compared with MAX-SMT:
Channel Capacity $=\log _{2}$ \#divisions

$$
\mathcal{H} \leq C C
$$

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$$
F_{1}\left(I, I_{1}, I_{2}\right)= \begin{cases}6 & : I>6 \wedge I_{1}>6 \\ I-1 & : 1 \leq I \leq 6 \wedge I \leq I_{1} \\ I_{1}-1 & : 1 \leq I_{1} \leq 6 \wedge I_{1}<I\end{cases}
$$

$F_{1}(\bar{L})$ tells you the size of the partition cell for $C_{1}$, for given $\bar{L}$.

## Maximizing Shannon Entropy Numerically

| $C_{1}=h<I \wedge h<I_{1}$ | $F_{1}(\bar{I})=\left\{\begin{array}{ll\|}8 & : I>8 \wedge I_{1}>8 \\ I-1 & : 1 \leq I \leq 8 \wedge I \leq I_{1} \\ I_{1}-1 & : 1 \leq I_{1} \leq 8 \wedge I_{1}<I\end{array}\right.$ |
| :--- | :--- |
| $C_{2}=h<I \wedge h \geq I_{1}$ | $F_{2}(\bar{I})= \begin{cases}8 & : I_{1}<1 \wedge 8<I \\ I-I_{1} & : 1 \leq I_{1} \leq I \leq 8 \\ I-1 & : I_{1}<1 \leq 1 \leq 8 \\ 9-I_{1} & : 1 \leq I_{1} \leq 8<1\end{cases}$ |
| $C_{3}=h \geq I \wedge h<I_{2}$ | $F_{3}(\bar{I})= \begin{cases}8 & : I<1 \wedge 8<I_{2} \\ I_{2}-I & : 1 \leq I \leq I_{2} \leq 8 \\ I_{2}-1 & : I<1 \leq I_{2} \leq 8 \\ 9-I & : 1 \leq I \leq 8<I_{2}\end{cases}$ |
| $C_{4}=h \geq I \wedge h \geq I_{2}$ | $F_{4}(\bar{I})= \begin{cases}8 & : I<1 \wedge I_{2}<1 \\ 9-I & 1 \leq I \leq 8 \wedge I_{2}<I \\ 9-I_{2} & : 1 \leq I_{2} \leq 8 \wedge I \leq I_{2}\end{cases}$ |

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$$
\frac{F_{1}(\bar{L})}{8}
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| :--- | :--- |
| $C_{2}=h<I \wedge h \geq I_{1}$ | $F_{2}(\bar{I})= \begin{cases}8 & : I_{1}<1 \wedge 8<I \\ I-I_{1} & : 1 \leq I_{1} \leq I \leq 8 \\ I-1 & : I_{1}<1 \leq I \leq 8 \\ 9-I_{1} & : 1 \leq I_{1} \leq 8<I\end{cases}$ |
| $C_{3}=h \geq I \wedge h<I_{2}$ | $F_{3}(\bar{I})= \begin{cases}8 & : I<1 \wedge 8<I_{2} \\ I_{2}-I & : 1 \leq I \leq I_{2} \leq 8 \\ I_{2}-1 & : I<1 \leq I_{2} \leq 8 \\ 9-I & : 1 \leq I \leq 8<I_{2}\end{cases}$ |
| $C_{4}=h \geq I \wedge h \geq I_{2}$ | $F_{4}(\bar{I})= \begin{cases}8 & : I<1 \wedge I_{2}<1 \\ 9-I & : 1 \leq I \leq 8 \wedge I_{2}<I \\ 9-I_{2} & : 1 \leq I_{2} \leq 8 \wedge I \leq I_{2}\end{cases}$ |

$$
\mathcal{H}(\bar{L})=\frac{F_{1}(\bar{L})}{8}
$$

## Maximizing Shannon Entropy Numerically

| $C_{1}=h<I \wedge h<I_{1}$ | $F_{1}(\bar{I})=\left\{\begin{array}{ll\|}8 & : I>8 \wedge I_{1}>8 \\ I-1 & : 1 \leq I \leq 8 \wedge I \leq I_{1} \\ I_{1}-1 & : 1 \leq I_{1} \leq 8 \wedge I_{1}<I\end{array}\right.$ |
| :--- | :--- |
| $C_{2}=h<I \wedge h \geq I_{1}$ | $F_{2}(\bar{l})= \begin{cases}8 & : I_{1}<1 \wedge 8<I \\ I-I_{1} & : 1 \leq I_{1} \leq I \leq 8 \\ I-1 & : I_{1}<1 \leq I \leq 8 \\ 9-I_{1} & : 1 \leq I_{1} \leq 8<I\end{cases}$ |
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\mathcal{H}(\bar{L})=\frac{F_{1}(\bar{L})}{8} \log _{2} \frac{8}{F_{1}(\bar{L})}+\frac{F_{2}(\bar{L})}{8} \log _{2} \frac{8}{F_{2}(\bar{L})}+\frac{F_{3}(\bar{L})}{8} \log _{2} \frac{8}{F_{3}(\bar{L})}+\frac{F_{4}(\bar{L})}{8} \log _{2} \frac{8}{F_{4}(\bar{L})}
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Numerically maximize $H(\bar{L})$

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\bar{L}=\langle 4,2,6\rangle
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$$

Numerically maximize $H(\bar{L})$

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First two steps of optimal binary search attack on 8 secrets.

# Finding Best Attack Tree Method 3 

Maximizing Shannon Entropy, Third Approach

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Maximum Satisfiable Subsets (MSS).
Optimization version of SAT.

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MSS solution $\Rightarrow$ maximize Shannon entropy.

## Finding Best Attack Tree

## Finding Best Attack Tree 3 Methods

# Finding Best Attack Tree 3 Methods 

## Do they work?

# Finding Best Attack Tree 3 Methods 

## Do they work?

Yes

## Implementation

- Java Symbolic Pathfinder (JPF / SPF) for symbolic execution.
- Specialized listeners for tracking observables (time, space).
- Latte and Barvinok for model counting path constraints.
- Max-SMT (Z3), MARCO (java + Z3) MSS.
- Mathematica's NMAXIMIZE for numeric maximization.
- Heuristics: top-down greedy optimization.


## Case study: Law Enforcement Employment Database

From DARPA Space-Time Analysis for Cybersecurity (STAC)

## Server

- 41 classes, 2844 line of code.
- stores all employee records by ID in a database.
- Some employee IDs have restricted access.


## Client

Commands available for users: SEARCH, INSERT, GET, PUT, ...
SEARCH a b has a timing channel: adaptive range query attack.

## Case study: Law Enforcement Employment Database

Domain: 100 possible IDs in database ( 6.541 bits)

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## Max SAT Subsets

- Attack tree depth: 7 (complete attack)
- Running time: 2 m 36 s


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## Numeric Entropy Maximization

- Attack tree depth: 11
- Incomplete attack: leaks 10.0 out of 19.9 bits
- Running time: 15m 8s


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## Max SAT Subsets

Does not scale to this domain.

## More Case Studies

We synthesized attacks for:

- ModPow used in RSA
- Compression Ratio Information Leak Made Easy (CRIME)
- java.util.Arrays.equal() (segment oracle attack)


## Conclusions

- Symbolic exection of adversary model to get constraint tree.
- Solve optimization problem to get low inputs to maximize leakage: attack tree.
- MAX-SMT

Symbolic Model Counting + Numeric Maximization Max-SAT-Subsets

- Experimentally validated our approach.


## Questions?

Thank you.
$28 / 29$

