Using Scan Side Channel to Detect IP Theft

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Outline

• IP theft issue in SoC

- Reverse Engineering with Scan
- Junta Learning
- Clustering and Graph Completion
- The Test Case: BitCoin SHA-256
- Conclusions

IP Piracy

- Modern SoC development mode: global and distributed
- IP passes dozens of hands

• Issue of Trust

Preventing IP theft

- Watermarks allow identification without altering the function
	- State Machine Encoding
	- Constraints on physical layout
	- More…
	- Detection
	- Proof
- Forensic techniques – Direct detection

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Reverse Engineering of an ASIC

- Phase 1 Invasive $Physical \rightarrow Circuit$
	- Delayering
	- SEM
	- Nanoscale Imaging
	- Cross-section

- Phase 2 Algorithmic $Circuit \rightarrow Spec$
	- FSM Extraction
	- Model Checking

– SAT

Reverse Engineering of an ASIC

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- Phase $2 -$ Algorithmic $Circuit \rightarrow Spec$
	- FSM Extraction
	- Model Checking
	- SAT Solvers

Scan Side Channel makes phase 1 non-invasive

Goal: automate production testing

Need to verify every net is functional

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Scan Insertion

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Shift Out

Unfolding Sequential Circuits with Scan

- Scan turns the SoC to a stateless circuit
- Mapped to the Boolean Function Learning problem: $\{0,1\}^n \rightarrow \{0,1\}^n$

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- Exponential Size: 2ⁿ

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Limited Transitive Fan-in

• In practice, logic cones have limited number of inputs: Transitive Fan $In = K$

Dependency Graph

Flip-flop Outputs

Flip-flop Inputs

- Bipartite graph represents flip-flop dependencies
- The goal: Find dependencies
- Complexity: $2^n \rightarrow 2^k$: Scalable with the chip size $\frac{1}{20}$

The K-Junta Algorithm

The K-Junta Algorithm
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y = f(\vec{x}), \vec{x} = \{x_1, x_2, ..., x_i, x_{i+1}, ..., x_j, ..., x_n\}
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\nGenerate random queries $y = f(\vec{x})$

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\nTheorem 1

\nThe following

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Test Institute

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Partial Dependency Graph

Flip-flop Outputs

Flip-flop Inputs

- If k is too high \rightarrow Partial dependency graph
- Influence = sensitivity of a function to a variable
- K-Junta works for Influence >1/2^K

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The Adder Example

- Dependencies across many bits are not likely to appear
	- Influence too low
- Close neighbor dependencies are discovered
- Need to group all the nodes of the adder

SNN Clustering

- Shared Nearest Neighbors Clustering
	- Every pair of nodes with >threshold shared dependencies assigned to the same cluster

SNN Clustering

Flip-flop Outputs

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Enumeration of the Adder Nodes

- Sort outputs in a cluster by their fan-in
	- Sort inputs accordingly
- Handle the plateau by iterative enumeration
	- $-$ Higher order inputs feed higher order outputs $\frac{32}{32}$

Flip-flop Outputs

Flip-flop Inputs

- Assuming the learner is looking for an adder
- Add dependencies of output bit *i* on all input bits 0 to *i*.

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SHA-256 Structure

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Learning Strategy

- The implementation is not known in advance
- But there are building blocks inherent to SHA- 256
	- 7-way adder
	- 5-way adder
- We search for structures that look like adders

BitCoin SHA-256 Accelerator

- Open source design from opencores.org
- Performance oriented, heavily pipelined
- ~80,000 registers
- Used a software simulator

After K-Junta and Clustering

Number of stages suggests two SHA-256 instances, but not necessarily

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Zooming in into a cluster

Detecting operands by fanout

W_i

- Fanout components \parallel
	- Bit order
	- bit oruer
— Number of functions $\begin{bmatrix} \frac{2}{5} \\ \frac{2}{5} \end{bmatrix}$
	- Function type

 E_i

F.

 G_i

 H_i

 C_i

D_i

 B_i

Returning to sequential

Flip-flop Inputs

Flattened

Summary

- A novel method of IP theft detection
	- By non-invasive reverse engineering with scan
	- Boolean function analysis and graph methods
	- Works with or without watermarks
- Learned a 80,000-register SHA-256 accelerator
- What next
	- More test cases
	- Detecting Trojan hardware

Thanks!

