Overview of Algorithmic Methods in IC Reverse Engineering

Leonid Azriel

Technion

1

- IC Reverse Engineering in a nutshell
- Scan-based netlist extraction
- Partitioning
- The matching problem
- Structural Analysis
- Functional Analysis
- Summary and Future Directions

Ethics of Reverse Engineering

- Semiconductor Chip Protection Act (US) 1984 allows reverse engineering of commercial semiconductor products for **educational purposes**
- Ethical goals
	- Detect vulnerabilities that allow reverse engineering and propose countermeasures
	- Explore legitimate usages of the reverse engineering for defense
		- Hardware Trojans detection
		- IP theft detection

Reverse Engineering of an IC

Phase 1 IC → Circuit

Phase 2 Circuit ➔ Spec

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Netlist Extraction

Hiroshi Fujiwara **Cyber Security Research Center**

Ender et al, Insights into the Mind of a Trojan Designer. The Challenge to Integrate a Trojan into the Bitstream," ASP-DAC 2019

101010 Tester Production

¹¹ Azriel, "Revealing On-chip Proprietary Security Functions with Scan Side Channel Based Reverse Engineering," *GLSVLSI 2017*

Production Tester

¹² Azriel, "Revealing On-chip Proprietary Security Functions with Scan Side Channel Based Reverse Engineering," *GLSVLSI 2017*

Production Tester

Unfolding Sequential Circuits with Scan

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

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- Exhaustive Search: Extract the Truth Table by running queries for all inputs

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- Mapped to the Boolean Function Learning problem: $\{0,1\}^n \rightarrow \{0,1\}^n$
- Exhaustive Search: Extract the Truth Table by running queries for all inputs
- Exponential Size: 2ⁿ

Shannon Effect

• Shannon Effect: "almost all" Boolean functions have a complexity close to the maximal possible $($ ^o $O(2^n))$ for the uniform probability distribution

• Corollary: For large n, "almost all" Boolean functions are not realizable in VLSI technology

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

Dependency Graph

Flip-flop Inputs

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

The K-Junta Algorithm $rac{1}{\overline{x},\overline{x}}$

Junta Algorithm	Hiroshi Fujiwar
$y = f(\vec{x}), \vec{x} = \{x_1, x_2, \ldots, x_i, x_{i+1} \ldots, x_j, \ldots, x_n\}$	

The K-Junta Algorithm $rac{1}{\overline{x},\overline{x}}$

Junta Algorithm

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y = f(\vec{x}), \vec{x} = \{x_1, x_2, \ldots, x_i, x_{i+1}, \ldots, x_j, \ldots, x_n\}
$$
\n**Exercise 2.1.1**

\n**Exercise 3.1.1**

\n**Exercise 4.1.1**

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\n**Exercise 5.1.1**

\n**Exercise 6.1.1**

\n**Exercise 7.1.1**

\n**Exercise 8.1.1**

\n**Exercise 9.1.1**

\n**Exercise 1.1.1**

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\text{ Generate random queries } y = f(\vec{x})
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The K-Junta Algorithm \overrightarrow{x} , \overrightarrow{x} =

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$\vec{a} = \{0, 0, \ldots 0, 0, 0, 0, 0, \ldots, 0, 0\}, f(\vec{a}) = 0$	
Generate random queries	$y = f(\vec{x})$
$\vec{b} = \{1, 0, \ldots 1, 0, 1, 0, 0, \ldots, 0, 1\}, f(\vec{b}) = 1$	

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The K-Junta Algorithm r r Algorith
 $(\vec{x}), \vec{x} = \{x_1, x_2$

....0,0,0,0,0,...

...0,0,0,0,0,...

Junta Algorithm

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y = f(\vec{x}), \vec{x} = \{x_1, x_2, \ldots, x_i, x_{i+1} \ldots, x_j, \ldots, x_n\}
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\n**ii** The identity of the *Research Center* is the *Research Center*.

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The K-Junta Algorithm r r

1unta Algorithm

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\nRelevant Variable

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Q(n \cdot \log n \cdot k \cdot 2^k)
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O\left(n \cdot \log n \cdot \mathrm{k} \cdot 2^k\right)
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$\vec{b} = \{1, 0, \ldots 1, 0, 1, 0, 0, \ldots, 0, 1\}, f(\vec{b}) = 1$	$Relevant Variable$

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Assumptions

- Physical access to the device
- Scan operation protocol is known
	- Vendors use proprietary protocols
	- There are standard solutions by EDA vendors
- Scan control is not protected or protection is defeated

• Or self-owned device

Assumptions

- Physical access to the device
- Scan operation protocol is known
	- Vendors use proprietary protocols
	- There are standard solutions by EDA vendors
- Scan control is not protected or protection is defeated

Scan-based netlist extraction setup

Fan-in analysis

- Partial data: not all dependencies discovered
	- Estimation: vast majority is there
	- Longer runs reveal few more links

• ~95% of logic cones can be directly reverse engineered at the logic function level

The two phases

Specification Discovery

• What actually the spec is?

• No universal definition or format

- Can be
	- Finding familiar elements
	- Translating to a higher level language
	- Goal dependent, e.g. finding specific properties

- IC Reverse Engineering in a nutshell
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- Partitioning
- The matching problem
- Structural Analysis
- Functional Analysis
- Summary and Future Directions

Circuit Partitioning

• Find basic structure in the flat graph
Circuit Partitioning

• Goal: match the source design partitioning

Circuit Partitioning

- Goal: match the source design partitioning
- Criteria: wire density

Circuit Partitioning

- Goal: match the source design partitioning
- Criteria: wire density
- Method: graph clustering

Minimum cut

- Partition a graph by removing minimum number of edges
- Widely used in the VLSI CAD tools
- Top-down approach
- Poor choice for our problem:
	- Fixed number of partitions
	- Uneven split

- Min-Ncut
	- Normalized min-cut

$$
NCut(A, B) = \frac{cut(A, B)}{assoc(A, V)} + \frac{cut(A, B)}{assoc(B, V)}.
$$

- Modularity maximization
	- Find best partition to maximize modularity

$$
Modularity(G_i \in G) = \frac{\left| E \right|_{G_i} - \left| E \right|_{G^*_{i} = \{V_{G_i}, E_{random}\}}}{\left| E \right|_{G}}
$$

• Every pair of nodes with **>threshold** shared neighbors are assigned to the same cluster

SNN (threshold=2)

Bottom-up

Clustering algorithms evaluation

• Goal: match the original design hierarchy

• Metric: Should reflect the goal

• NMI: Normalized Mutual Information – Information theoretical metric

UART

RISC-V tov model

 0.8 0.7

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Matching against a library

- Structural Analysis
	- Syntactic match
	- Isomorphism
	- Less flexible

- Functional Analysis
	- Semantic match
	- Functional identity
	- Static / simulations

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Matching library components

- Subgraph isomorphism
	- Structural match
	- NP-complete

- Heuristics
	- Dimension reduction
	- Signatures

NPNP-invariant matching

NPNP-invariant matching

• Input/output permutation

NPNP-invariant matching

- Input/output permutation
- Input/output negation

Technology mapping

- RTL synthesis
	- Map generic logic to technology cells
	- npnp-invariant matching

Outline

- IC Reverse Engineering in a nutshell
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Looking for more structure

WordRev: construct bit slices

6-Feasible cut (K-Feasible cut of $N = K$ nodes that completely define the value of N)

WordRev: group equivalent slices

Permutation-invariant Boolean matching

⁵⁴ Li, "WordRev: Finding word-level structures in a sea of bit-level gates," HOST 2013

WordRev: group by connections

Parallel connection (e.g. MUX select)

Serial connection (e.g. carry propagation)

WordRev: propagate word structures

DANA - Universal Dataflow Analysis for Gate-Level Netlist Reverse Engineering

• Detecting registers in data words

- Loose topological comparison of fan-in structures
- Preprocess to simple gates
- Recursive search

RELIC tool

- Similarity metrics
- Based on number of inputs
- Connecting nodes with *similarity > threshold*

- Mark registers with feedback path as FSM registers
	- Too many false positives

- Mark registers with feedback path as FSM registers
	- Too many false positives
- Narrowing down
	- Select registers that control datapath
	- Group registers that have the same enable signal
	- Group registers that have shared gates in their combinational feedback path

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- Massively used in the VLSI design process
- Uses anchors
	- Named registers
	- Hierarchical boundaries
- Usually combinational match only
	- Still a SAT problem, but there is a structure

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• Some unsupervised learning used for clustering

• Can the full ML power be harnessed for RE?

• Problem: insufficient dataset

- Map a circuit to a library of 4-LUTs
- 222 isomorphism classes to make the LUT npn-invariant
- Existence vectors: classes of adjacent elements
- Reduced existence matrix is fed to CNN
- Total set of 250 synthetic circuits
- Efficient for a small number of classes
	- One class (detect whether multiplier is present): 97% accuracy
	- 9 classes: 75% accuracy

Spectral Graph Methods

Reverse Engineering Tools

- Commercial
	- ChipWorks
	- Texplained ChipJuice (phase 1)
- Academia
	- Degate (phase 1)
	- HAL

Outline

- IC Reverse Engineering at a glimpse
- Partitioning
- The matching problem
- Structural Analysis
- Functional Analysis
- Algorithmic Reverse Engineering Tools
- Summary and Future Directions
- Uniform output format
	- Streamline the research
	- Evaluation baseline and common metrics
- Quantifying success
	- How to measure spec information
	- Is number of classified gates the right metric
- Graph methods
	- Take advantage of the big advancement in the social network analysis

- Functional analysis
	- Use Boolean function properties
- Machine learning
	- Generate a synthetic data set
	- Generative Adversarial Networks
	- One-shot learning
- Protection
	- Logical obfuscation
	- Structural obfuscation

Questions

Matching using simulation vectors

- Permutation invariance
	- One-hot and two-hot vectors
	- Simulation graphs
	- Matching subgraphs

Matching using simulation vectors

- One-hot and two-hot vectors
- Simulation graphs
- Matching subgraphs
- Negation invariance
	- Add polarity variables
	- SAT problem

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