HybMT: Hybrid Meta-Predictor based ML Algorithm for Fast Test Vector Generation



Shruti Pandey, Jayadeva and Smruti. R. Sarangi Department of Electrical Engineering Indian Institute of Technology Delhi New Delhi, India

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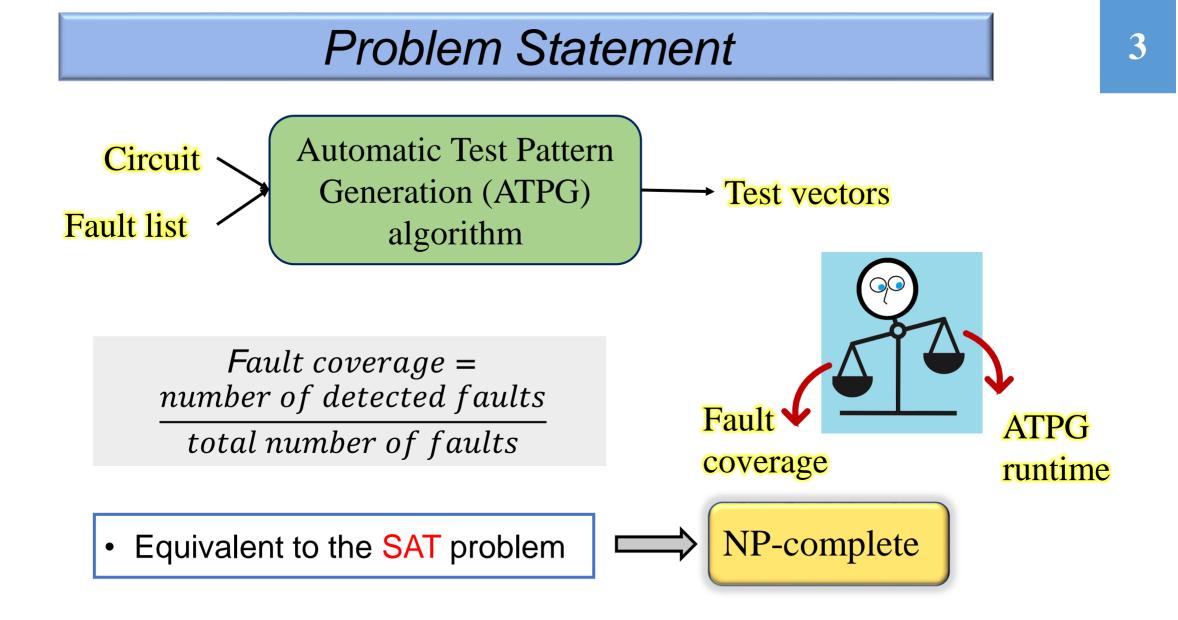
Outline

Introduction and Motivation

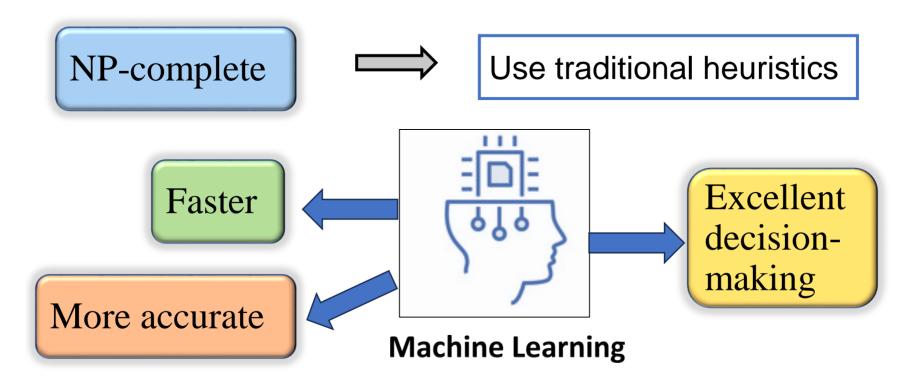
Background

Methodology

Results and Conclusion



Problem Statement





Minimize the test vector generation time using ML subject to sufficient fault coverage.



Have not moved beyond the **4-decade old ISCAS '85** benchmarks.

No comparisons with the **commercial tools**.

No work demonstrated **consistent improvement** across all the benchmark circuits.

HybMT Overcomes these Pitfalls



Generates test inputs for recently released benchmark circuits that are up to **70× larger** than the ISCAS'85 circuits.

Outperforms a popular commercial ATPG tool

Shows a consistent speedup across all the benchmark circuits.

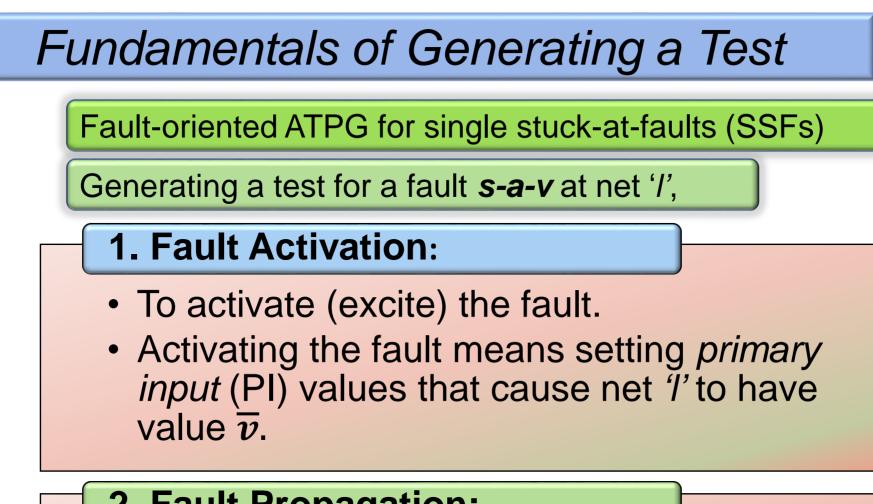
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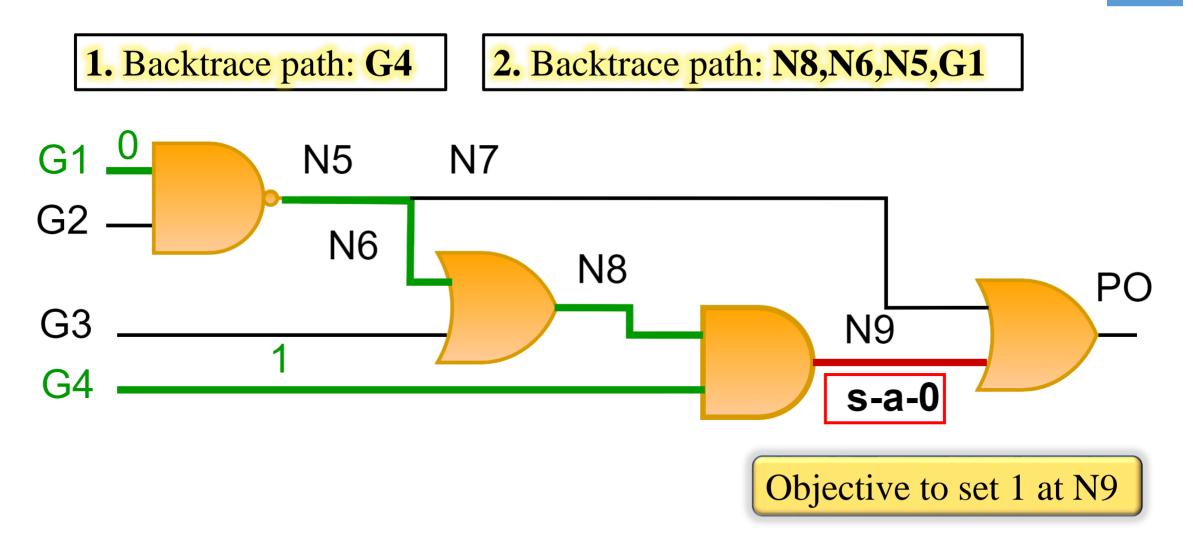
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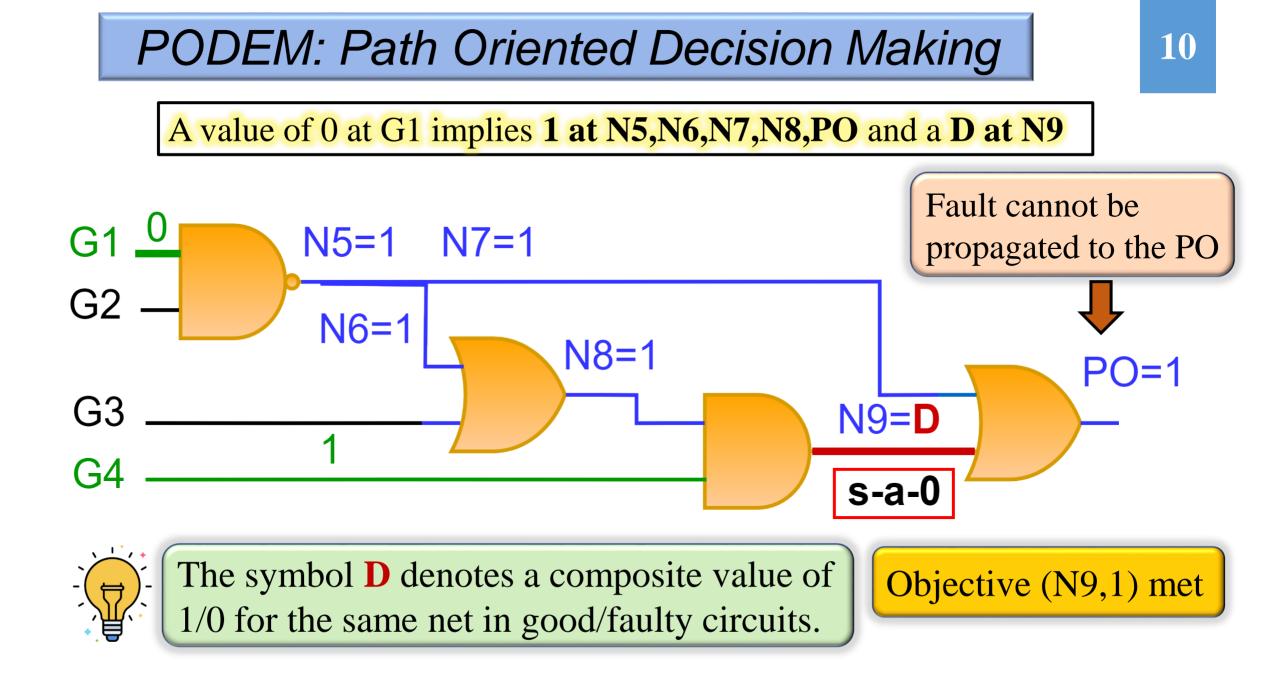


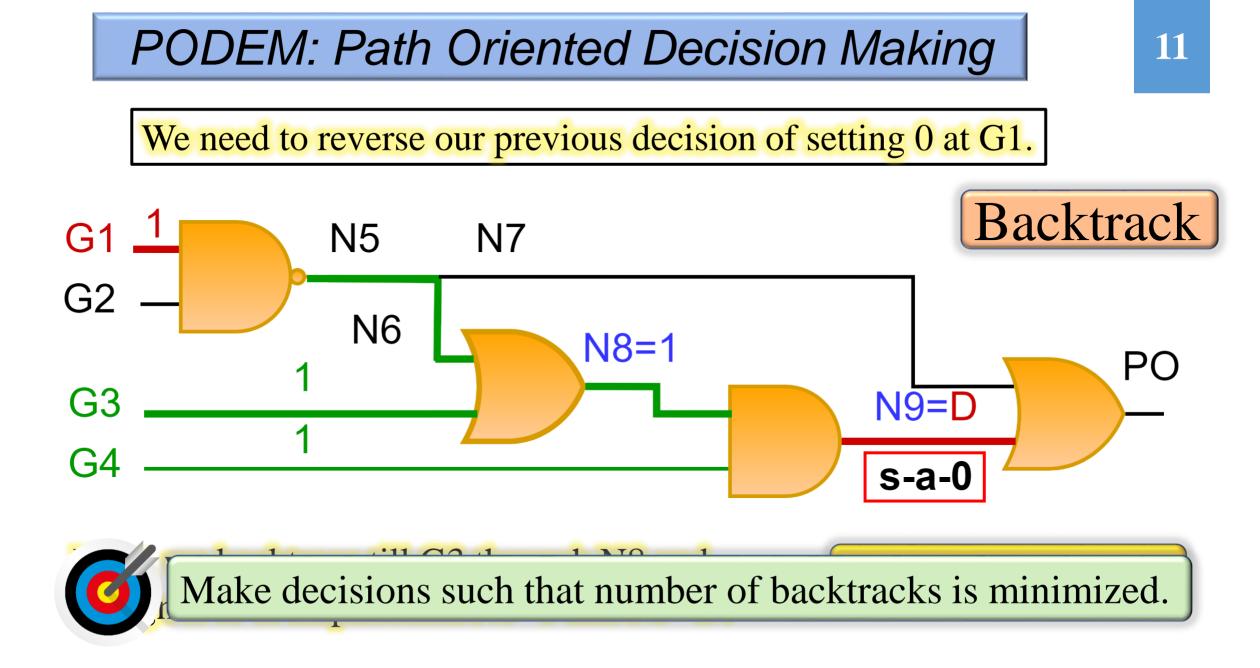
2. Fault Propagation:

• To propagate the fault to a *primary output* (PO).



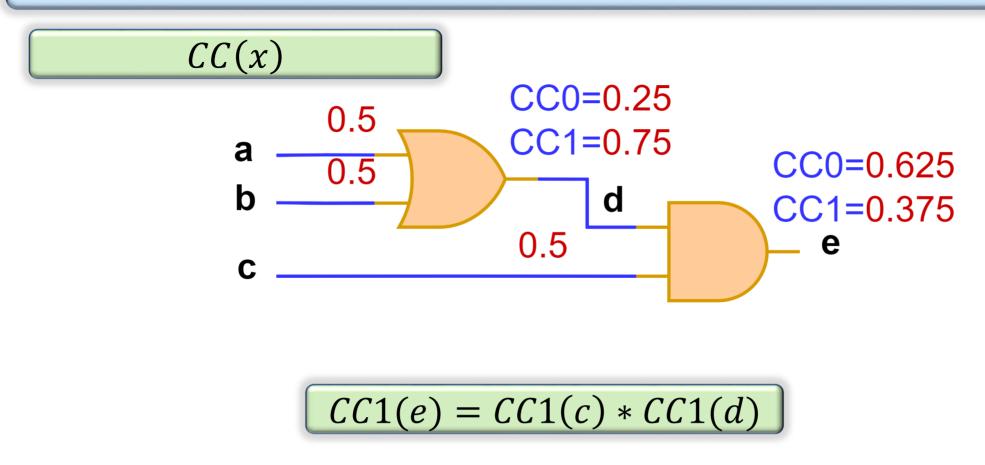






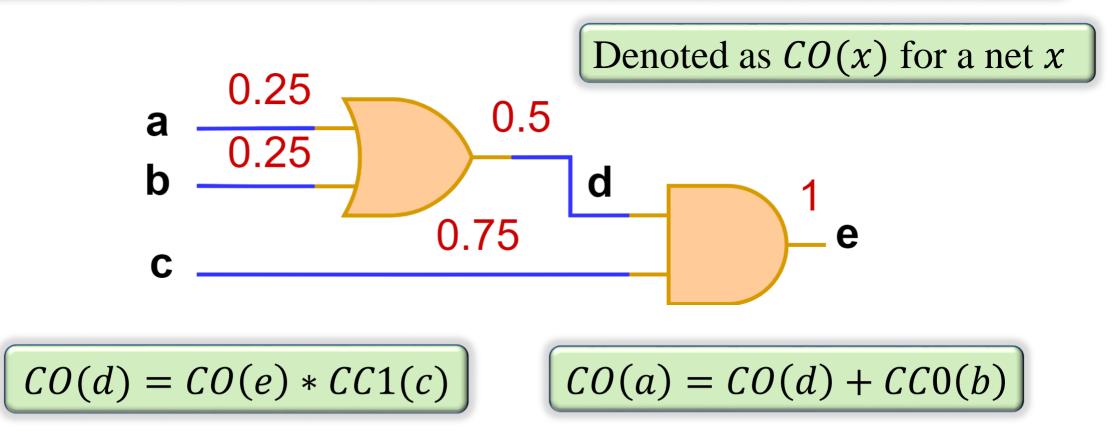
Controllability

The probability of setting a net to 0 or 1 by making random PI assignments



Observability

The probability of propagating the value at a net to a primary output (PO).



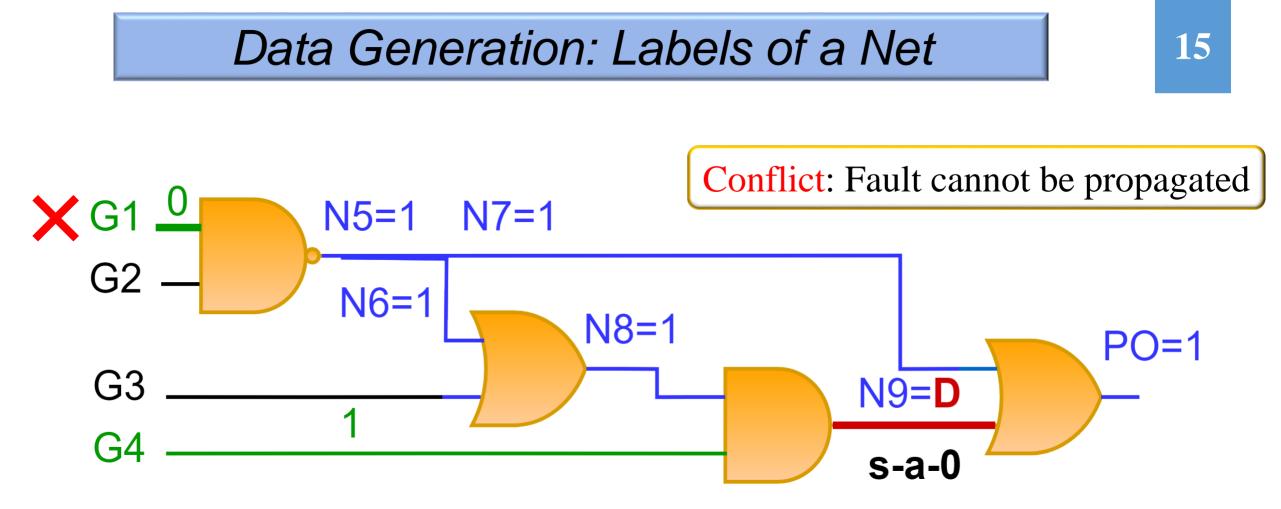
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If PI assignment leads to a conflict ground truth label for the nets in the backtrace path is 0 else 1

Data Generation: No-backtrack Probability

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- A circuit net occurs several times while running PODEM.
- Its input features remain the same,
- But the ground truth label keeps changing.

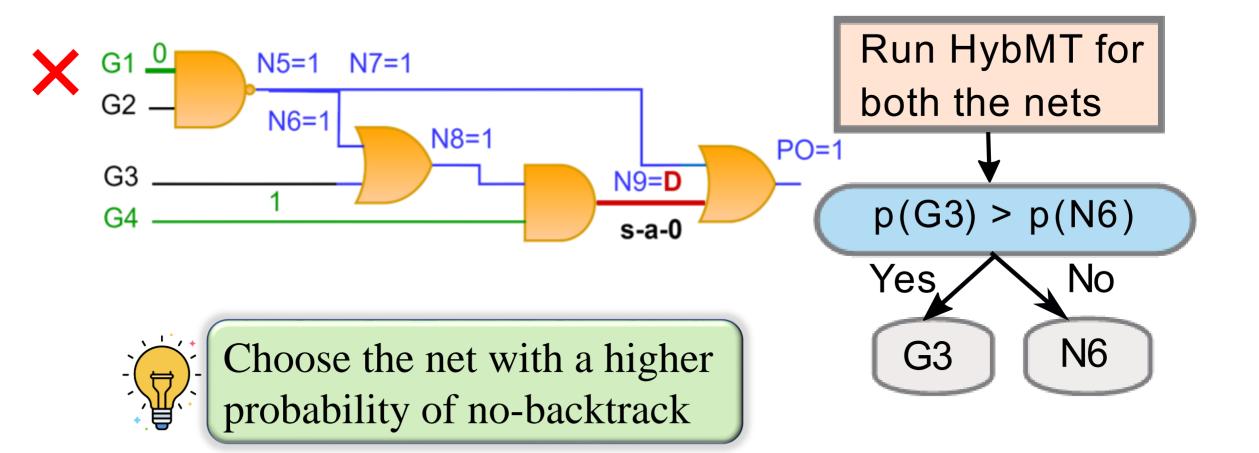


Probability of no-backtrack

(# ground truth label for circuit net 'l' is 1)

(frequency of occurrence of circuit net 'l')

HybMT based PODEM

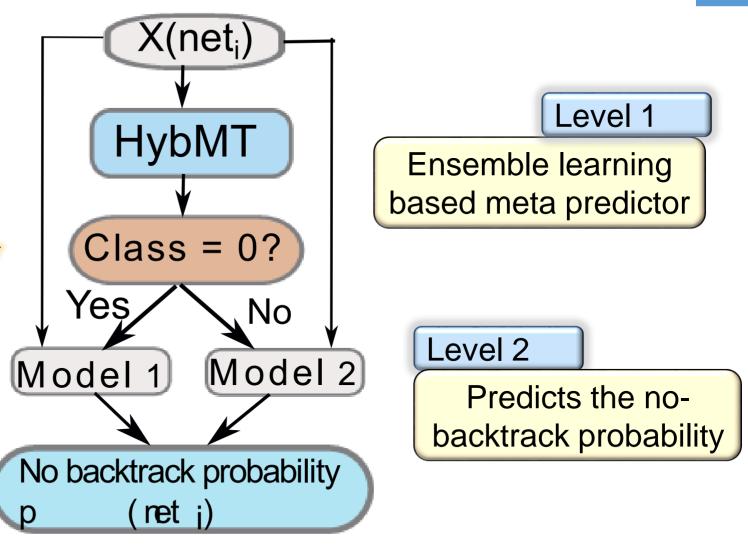


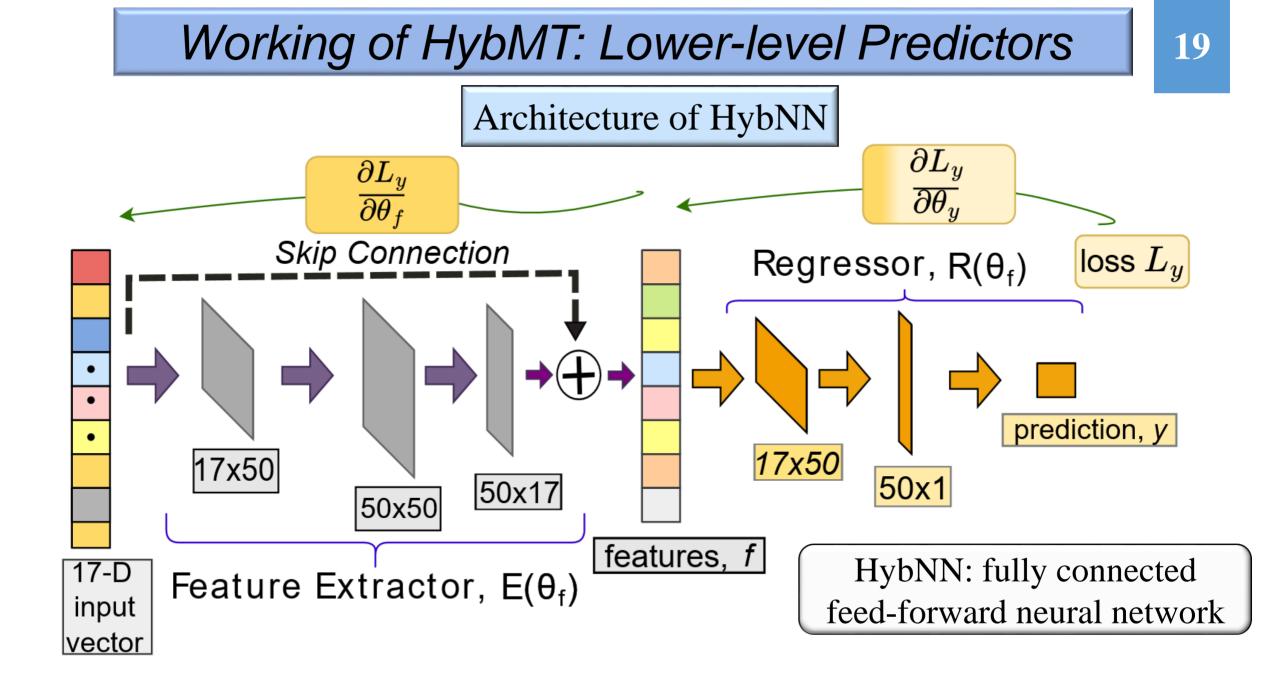
Working of HybMT

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 $X(net_i)$ represents the features of net_i

 $p(net_i)$ is the probability of no backtrack at net_i





Features of the ML Models

1. SCOAP controllability* Second-level **HybMT** 2. SCOAP observability** Predictor 3. Fanout of the gate *minimum number of PIs required to 1. COP controllability set the value of a net to a given value 2. COP observability **minimum number of PI Shortest distance from the primary input (PI) (normalized) assignments that need to be 4. Type of the gate (AND, OR etc.) made to propagate the fault to a PO

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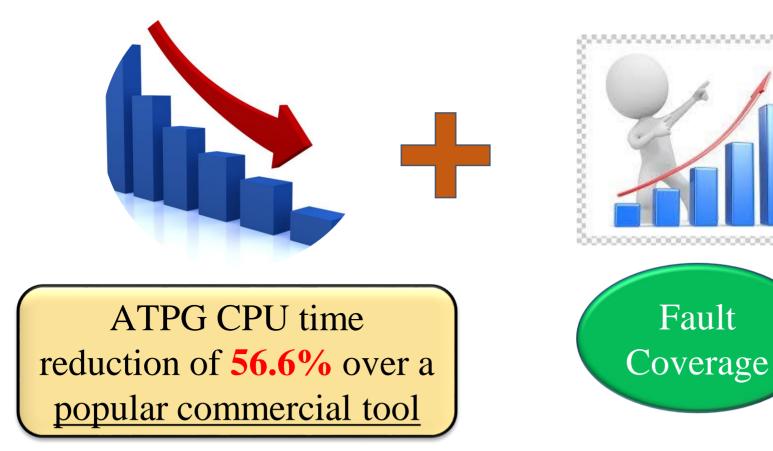
Modified the open-source ATPG framework for PODEM obtained from Fault, a part of Google's OpenLane project.

Loss function: Mean squared error.

Optimizer: Adam with a learning rate of 0.01.

Weight initializer: uniformly distributed in the range $(-\sqrt{k}, \sqrt{k})$ where $k = \frac{1}{in_features}$.

Results: EPFL Benchmark Circuits



ATPG CPU time reduction of **126.4%** over the best ML-based ATPG algorithm

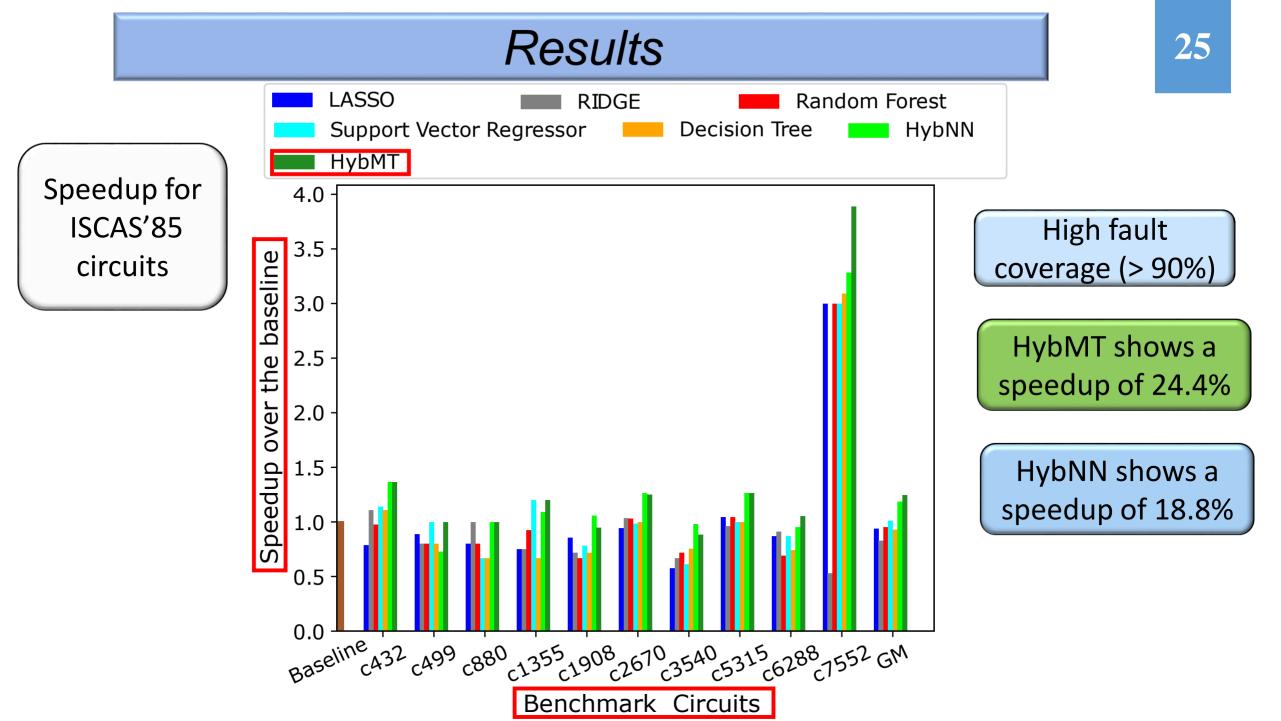
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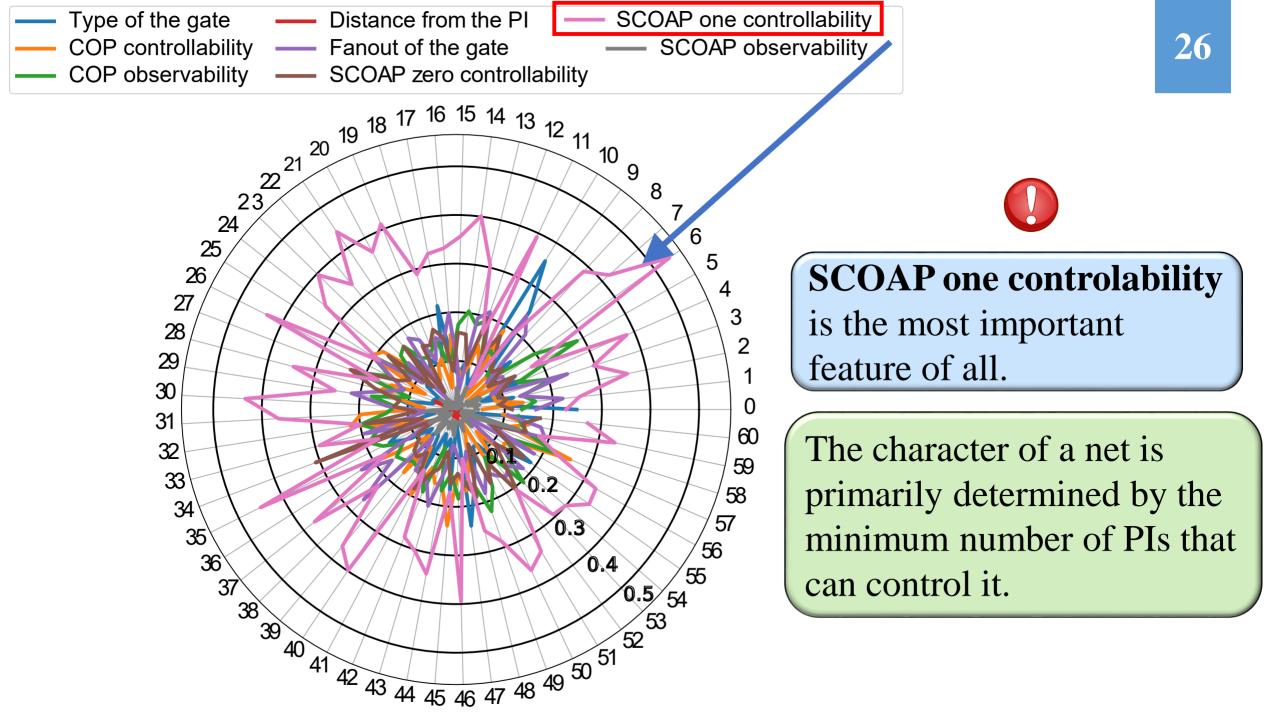
Results: CPU Time and Fault Coverage



First to beat a commercial tool !!

EPFL	CPU Time (in ms)			Fault Coverage		
Benchmark				(in %)		
Circuits	Commercial	Baseline	HybMT	Commercial	Baseline	HybMT
	ATPG Tool			ATPG Tool		
Adder	190	2,891	180	100	100	100
Barrel-shifter	340	50	56	100	100	100
Max	950	44	780	100	100	100
Multiplier	2,510	2,545	2,200	100	100	100
Sine	790	3,25,340	708	97	74	97
Square	1,870	647	800	98	98	98





Conclusion

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A two-level predictor *HybMT* for key decision making in PODEM.

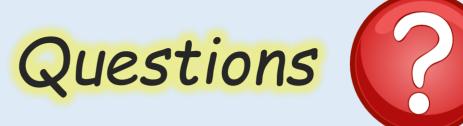
Reduces PODEM's CPU time without sacrificing on fault coverage.

Obtains a speedup of **56.6%** and **126.4%** over the commercial tool and the state-of-the-art approach respectively.

Properly designed neural networks like **HybNN** are better candidates for guiding backtracing.







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