



DEEPSECURITY: APPLYING DEEP LEARNING TO HARDWARE SECURITY



**QUEEN'S
UNIVERSITY
BELFAST**

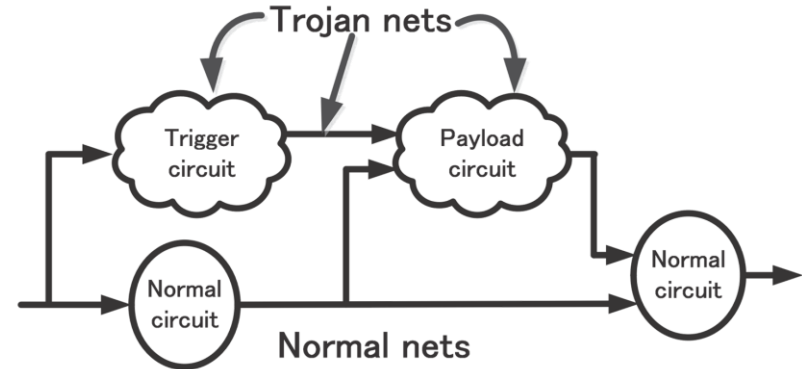
DeepSecurity: Applying Deep Learning to Hardware Security

Overall Goal

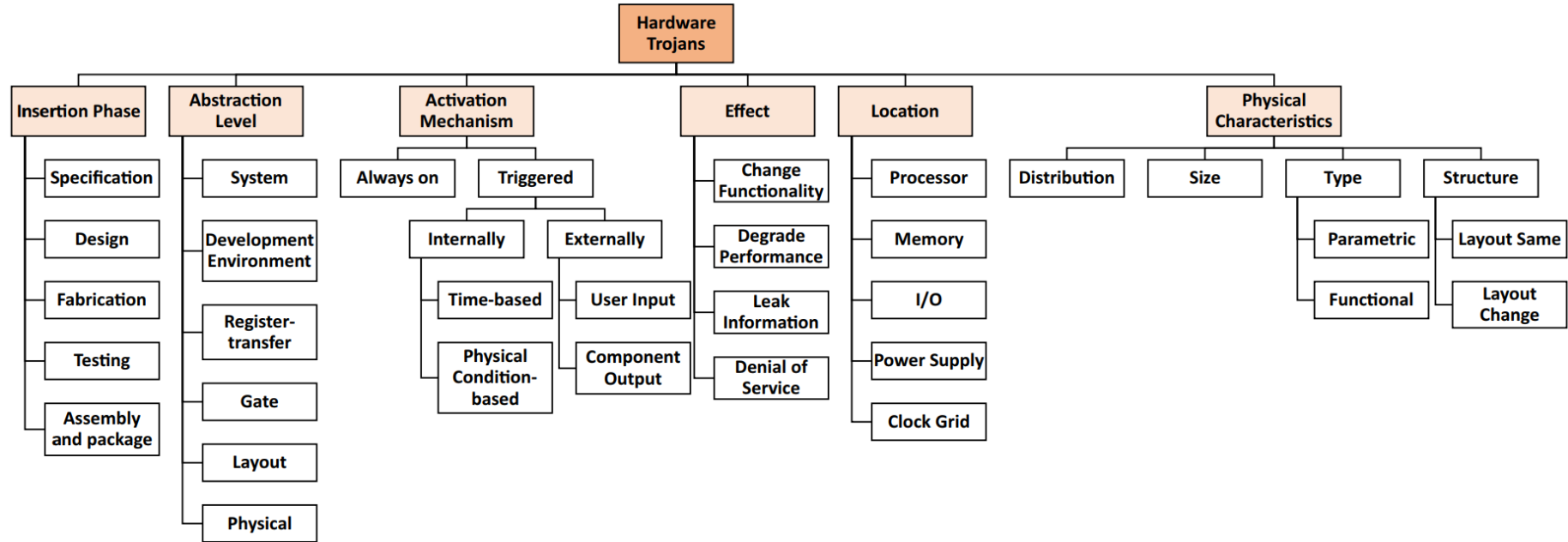
To investigate the use of Deep Learning for security verification in EDA tools, specifically in relation to *Hardware Trojan detection* and *Side channel analysis* to allow non-security experts to receive feedback on how to improve the security of their designs prior to fabrication.

Hardware Trojans

- Hardware Trojans (HTs) are malicious modifications to integrated circuits (ICs)
 - Emerging security concern in the IC industry
- Globalisation of semiconductor supply chains, design and fabrication of ICs now distributed worldwide
 - use of overseas foundries,
 - third party IP,
 - third party test facilities
- Becoming difficult to ensure the integrity and authenticity of devices



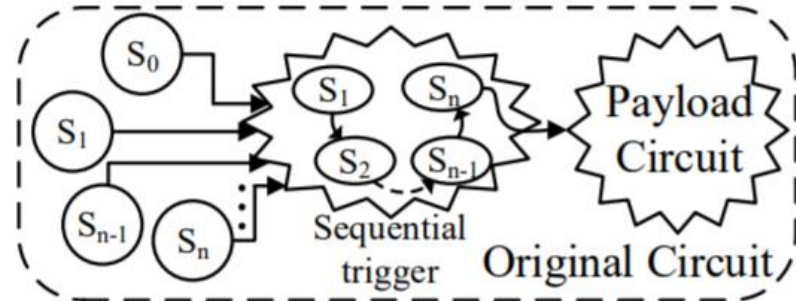
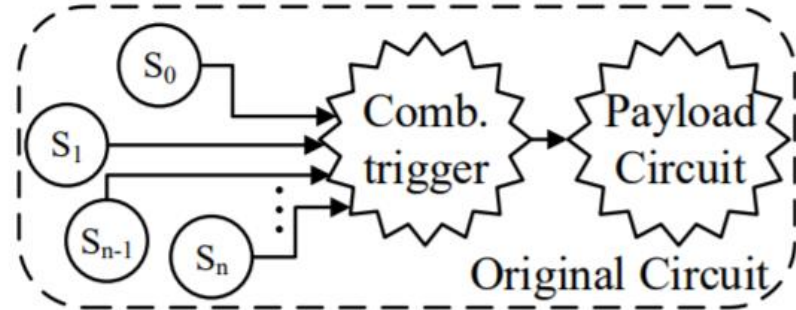
HT Classification



Source: B. Shakya, T. He, H. Salmani, D. Forte, S. Bhunia, and M. Tehranipoor, 'Benchmarking of Hardware Trojans and Maliciously Affected Circuits', *J. Hardw. Syst. Secur.*, pp. 1–18, 2017.

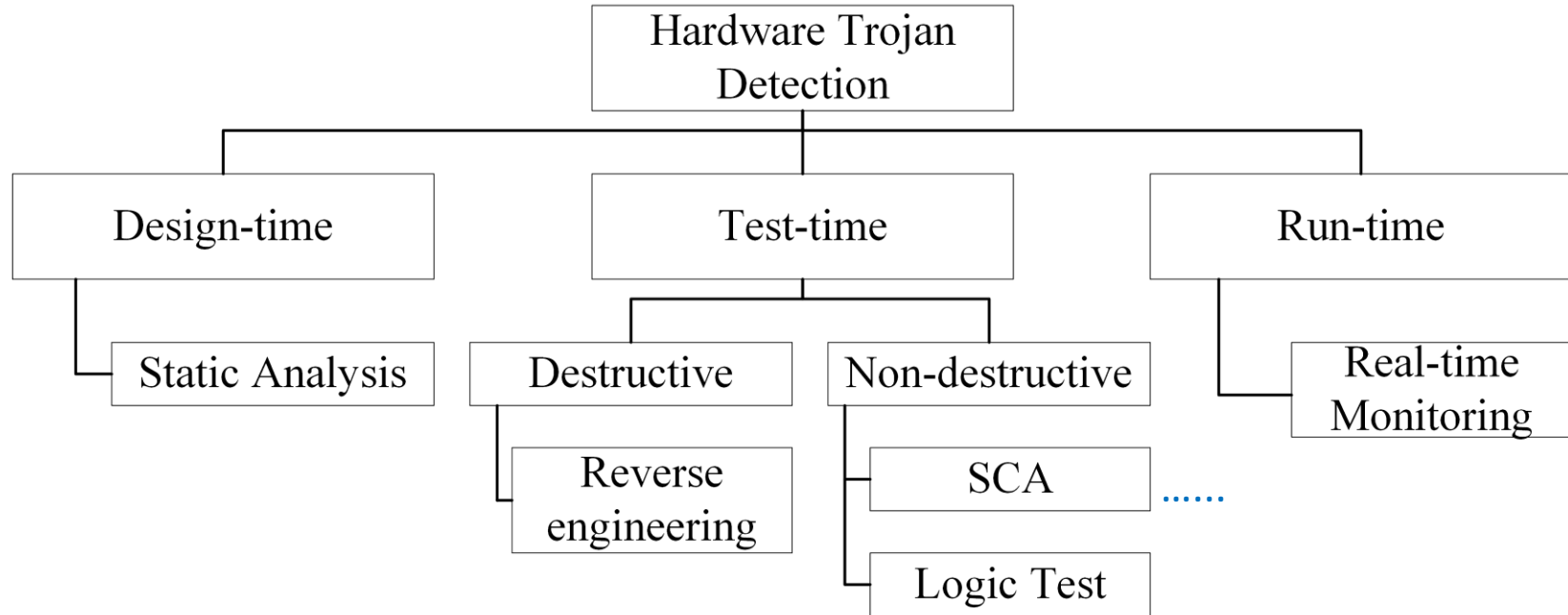
Simple Classification

- Functional Trojans
 - Combinational Trojans
 - Sequential Trojans
- Parametric Trojans
 - Dopant area
 - Doping concentration

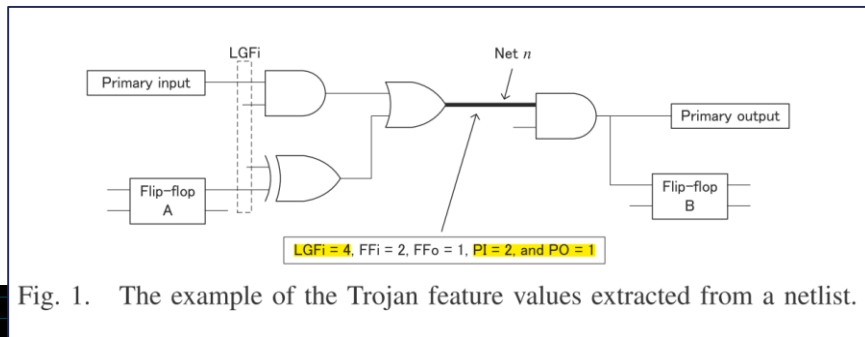


Detecting Hardware Trojans

– IC Production Stage



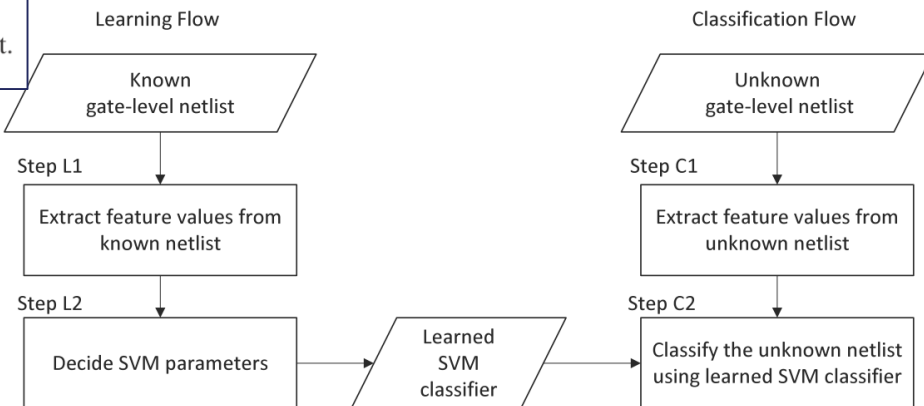
Use of Machine Learning to detect HTs



Data	TPR in [1]	TPR in ours	TNR in [1]	TNR in ours
s15850-T100	61%	93%	99%	66%
s35392-T200	27%	100%	99%	59%
s38417-T100	100%	100%	99%	76%
s38584-T200	99%	94%	98%	64%

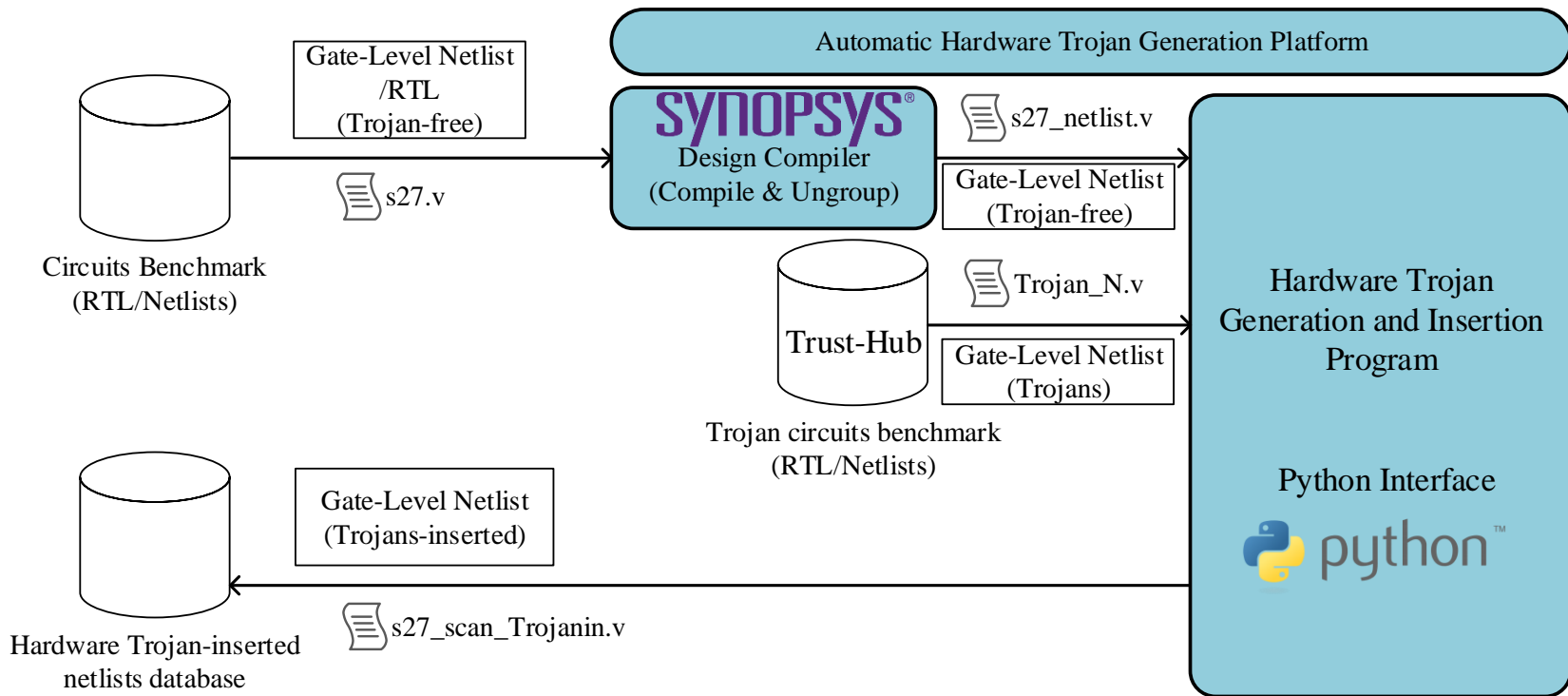
AVERAGE VALUES OF FFi, FFo, PI, AND PO.

Trojan/Normal	FFi	FFo	PI	PO
Trojan nets	1.11	1.51	4.60	3.65
Normal Nets	0.18	0.16	1.50	1.35



Source: K. Hasegawa, 'Hardware Trojans classification for gate-level netlists based on machine learning', IEEE 22nd IOLTS, 2016.

Current progress: An Automatic Hardware Trojan Generation Platform-1/2

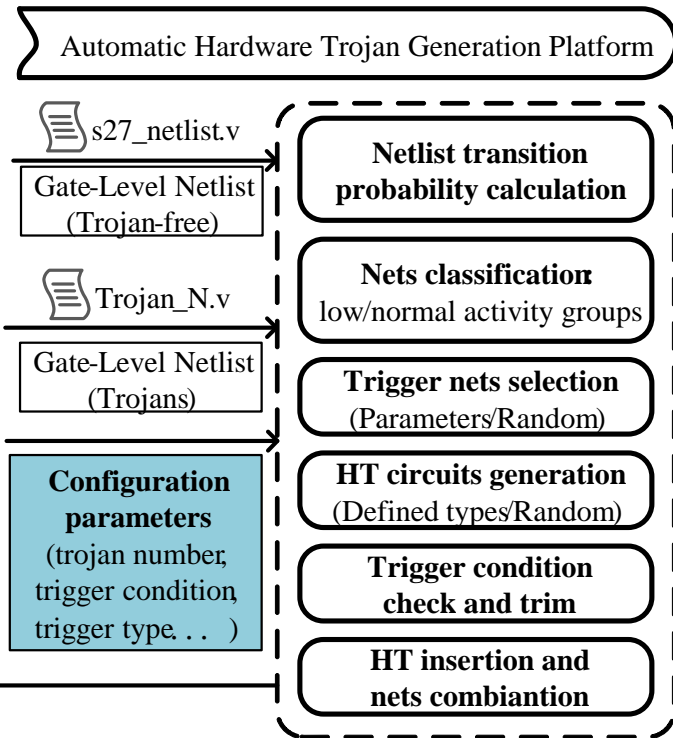


Current progress: An Automatic Hardware Trojan Generation Platform-2/2

Sample of HT Configuration Parameters

Parameters	Value Type
No. of Hardware Trojans	Number
No. of low activity nets	Number
No. of normal activity nets	Number
HT trigger type	Combinational/Sequential
HT payload type	Functional/Leakage
User-designed HT	Enable/Disable
.....

- Trigger selection according to netlist transition probability
- Highly configurable
- A variety of HT types and benchmark circuit choices
- Pre-research on ML/DL-based HT detection



Next Steps

- Evaluate ML approaches to check the validity of the proposed platform and HT-infected circuit database & make the database publicly available.
- Compare several ML algorithms to determine the best detection approach.
- Step into deep learning, investigate the performance between different NN architectures and build a novel deep learning architecture for HT detection.
- Conduct a comprehensive evaluation of the application of supervised and unsupervised ML and deep learning techniques for HT detection on gate-level netlists.



Thank you