

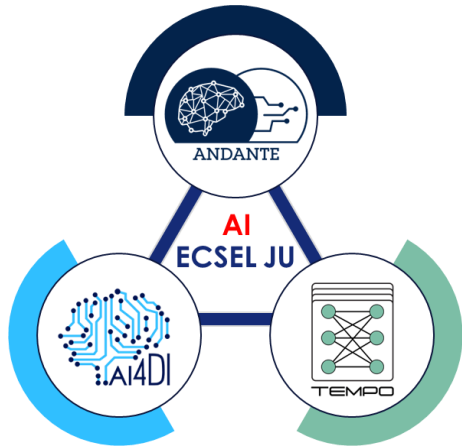


International Workshop on Embedded Artificial Intelligence Devices, Systems, and Industrial Applications (EAI)



Milan, Italy 19 September 2022

International Workshop on Embedded Artificial Intelligence Devices, Systems, and Industrial Applications (EAI)



Low-Power Vertically Stacked One Time Programmable Multi-bit IGZO-Based BEOL Compatible Ferroelectric TFT Memory Devices with Lifelong Retention for Monolithic 3D-Inference Engine Applications



Sourav De

19 September 2022 Milan, Italy

Presentation Outline



- Introduction
- Process flow
- Capacitor Characterization
- Transistor Characterization
- Implications to Neural Network
- Benchmarking

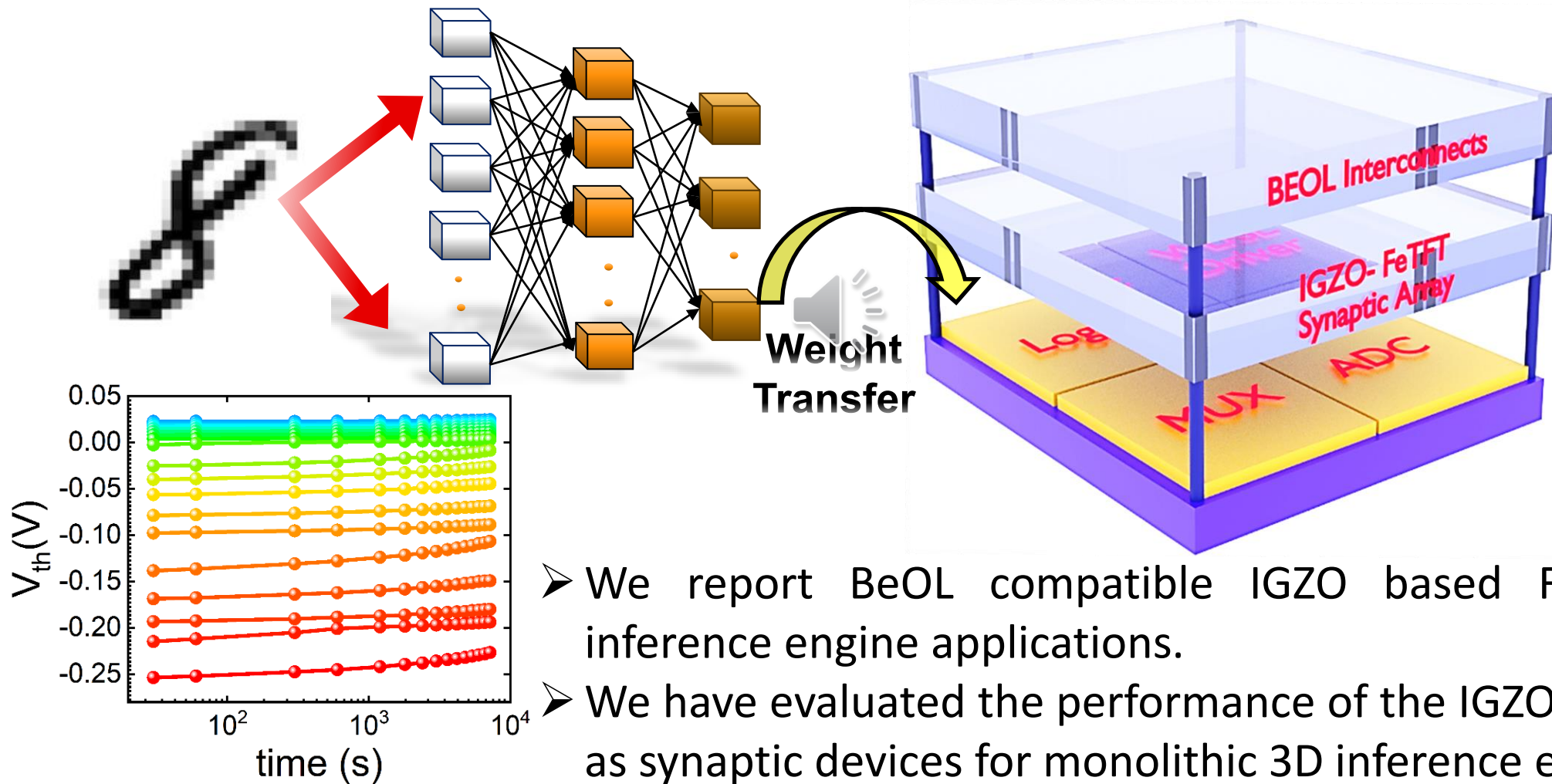
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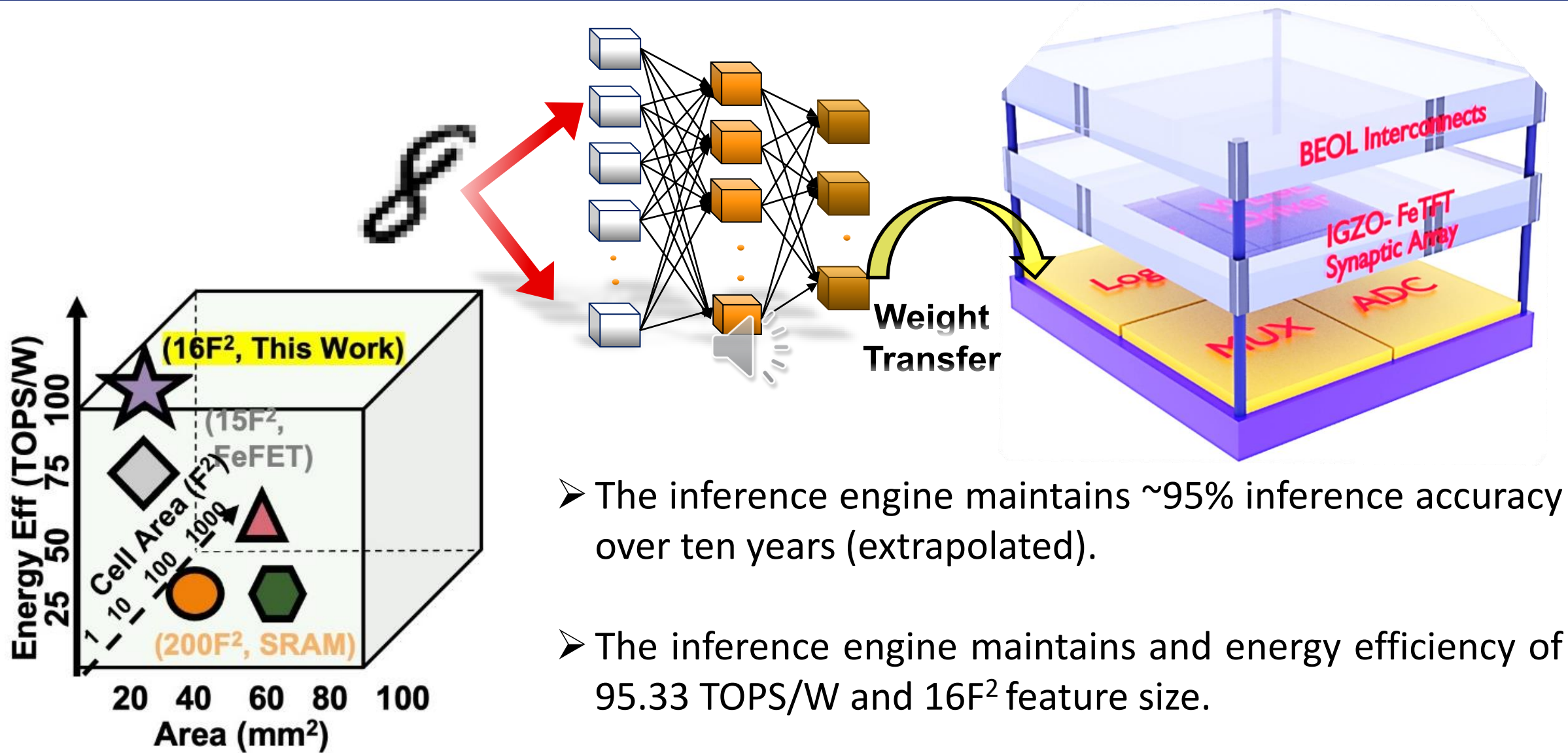
Introduction

MLP Neural Network



- We report BeOL compatible IGZO based FeTFT device inference engine applications.
- We have evaluated the performance of the IGZO-based FeTFT as synaptic devices for monolithic 3D inference engine.

Introduction

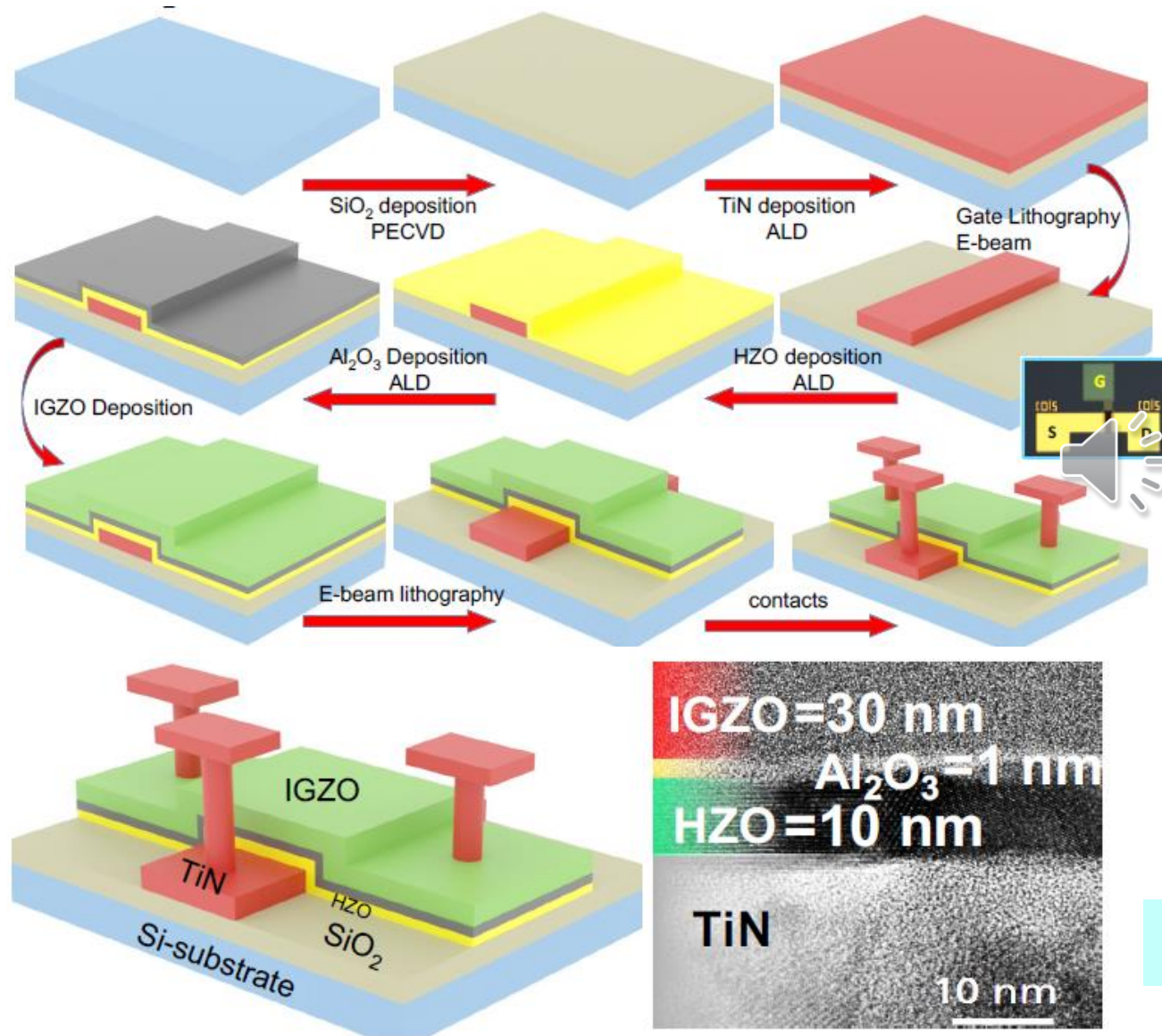


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Process flow



- Standard cleaning of Si wafer
- Deposition of SiO_2 by PECVD
- Deposition of TiN by ALD
- Gate lithography by E-beam
- Deposition of HZO by ALD
- Deposition of Al_2O_3 by ALD
- Deposition of IGZO
- E-beam lithography
- Contact pad deposition.

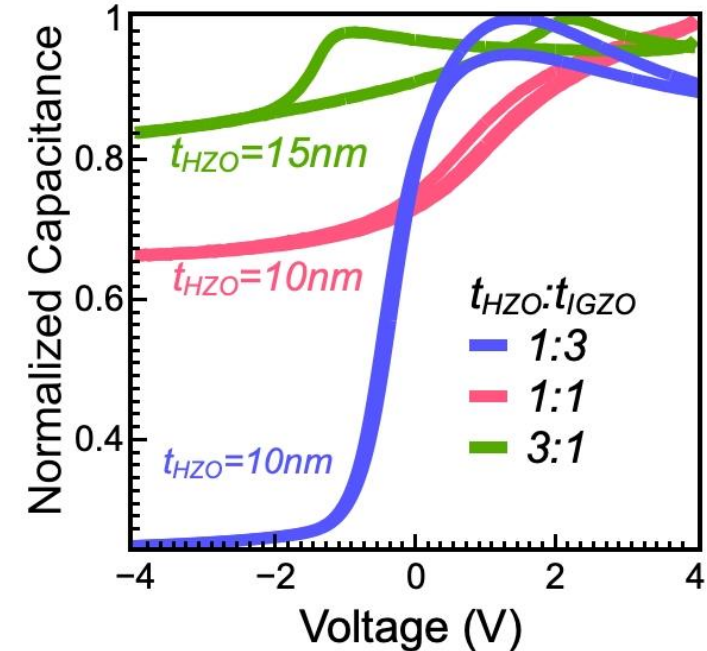
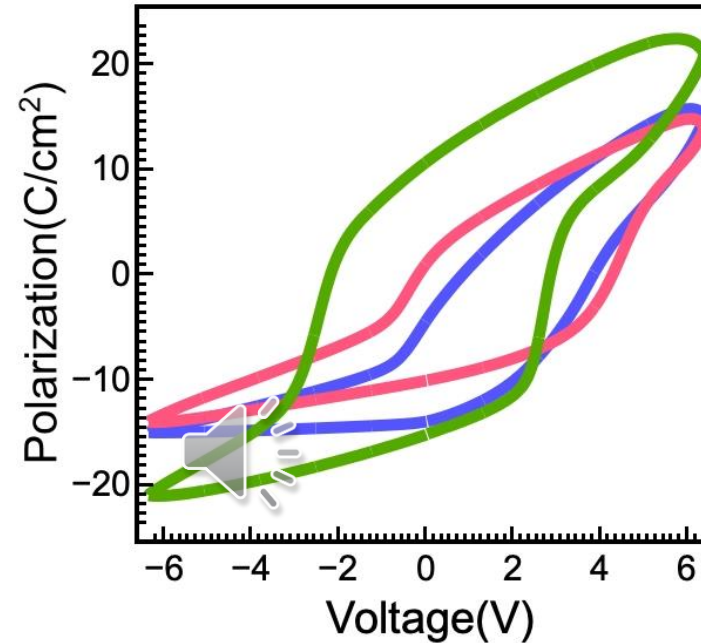
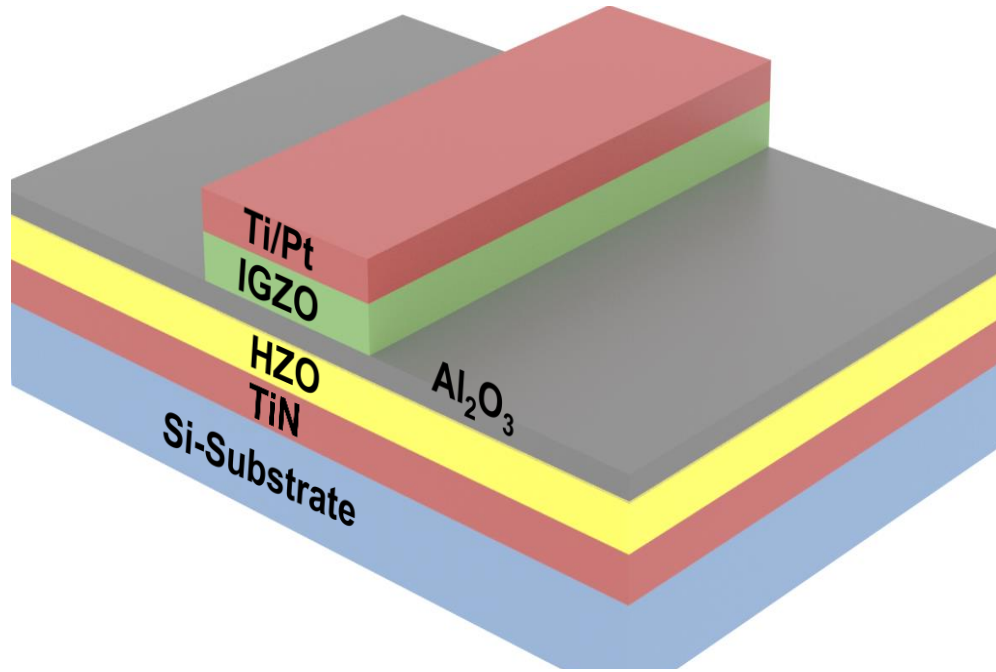
Maximum process temperature is 350°C

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Capacitor Characterization



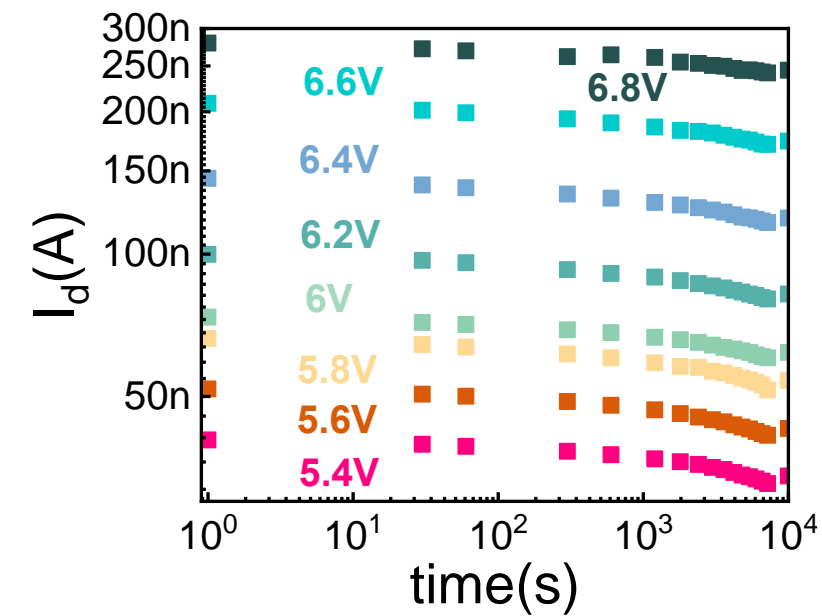
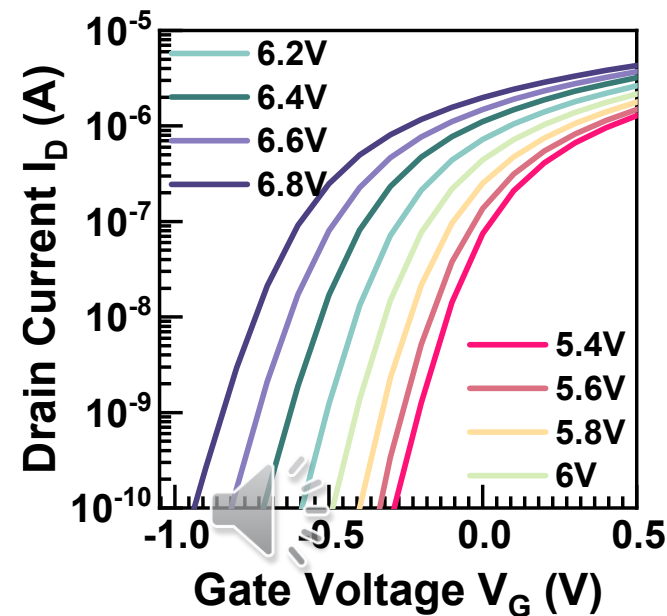
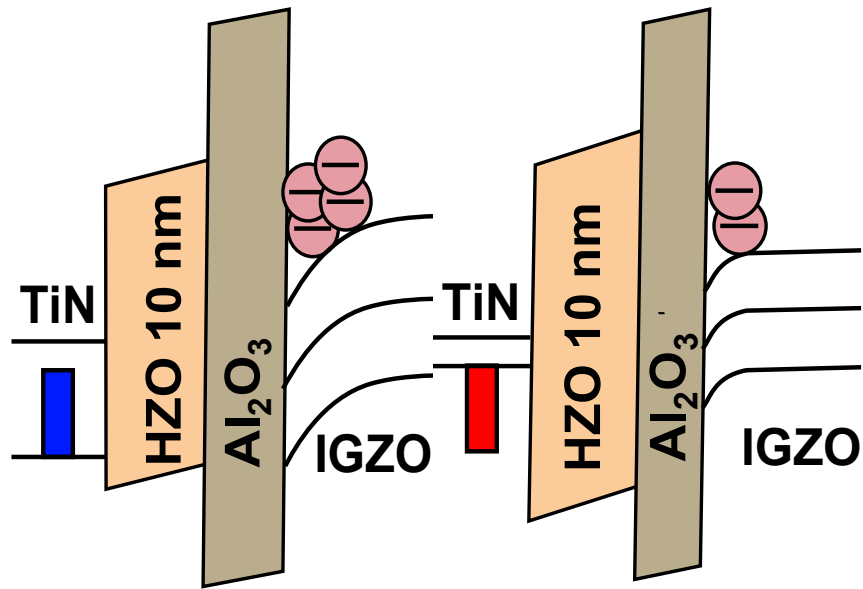
- Schematic of the gate-stack with IGZO as semiconductor and HZO as a ferroelectric layer.
- Polarization versus voltage (P-V) response of metal-semiconductor-FE-metal (MSFM) gate stacks with various thicknesses of the HZO and IGZO.
- Capacitance versus voltage (C-V) response of metal-semiconductor-FE-metal (MSFM) gate stacks with various thicknesses of the HZO and IGZO.

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Transistor Characterization



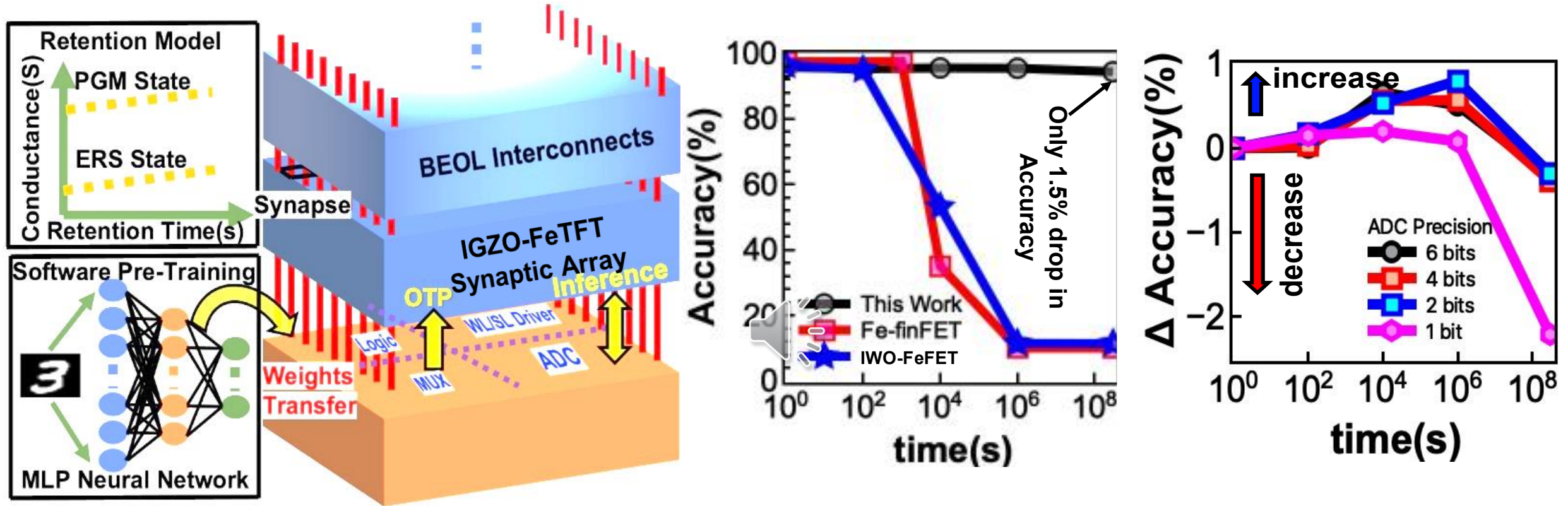
- The band diagram of the IGZO-MOS capacitor during the program and erase operation. Although the high value of μ_n facilitates fast programming, the low value of μ_p inhibits the erase operation by preventing the supply of holes during erase operation.
- 3 bits/ cell WRITE operation in IGZO-based FeTFTs with 200 ns wide pulses of minimum amplitude 5.2 V and maximum amplitude of 6.8 V.
- The measured retention characteristics show stable retention of 8-states for 10^4 seconds and for ten years for 2bits/cell operation with negligible degradation.

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Implications to Neural Network



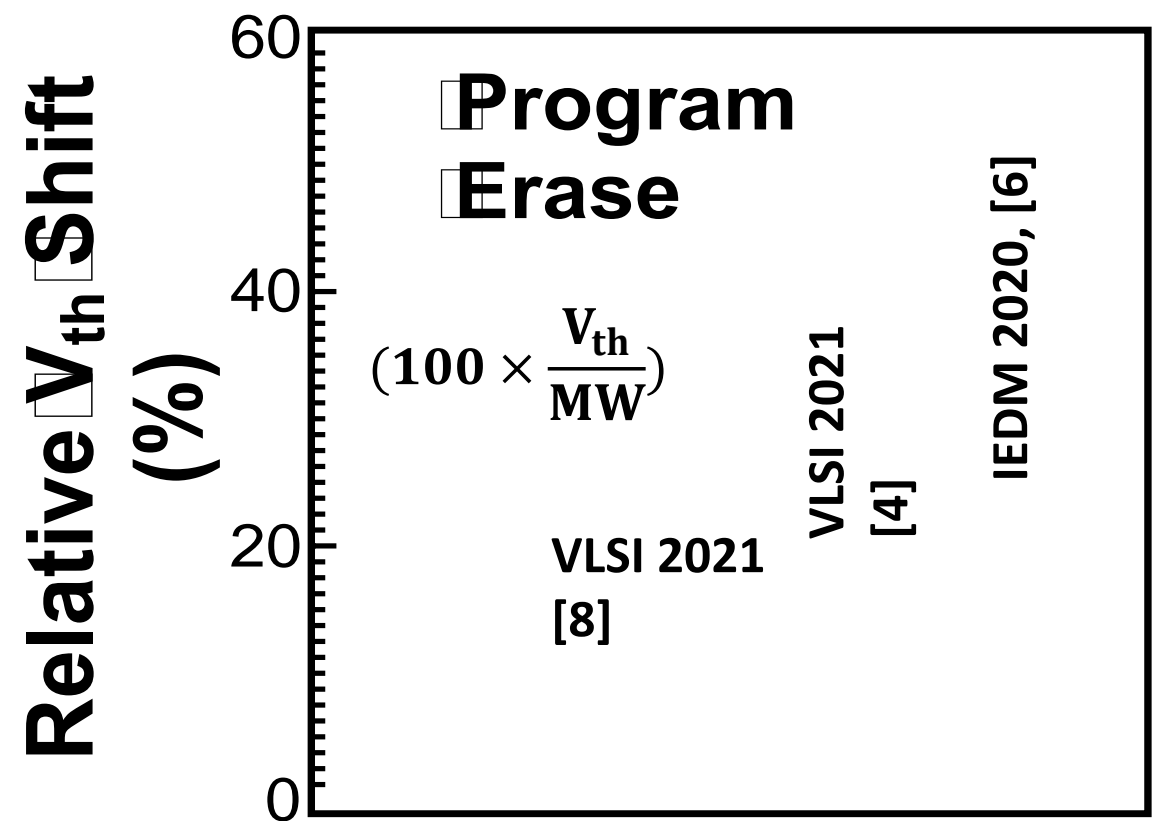
- The modus operandi of monolithic 3D inference engine based on multilayer perceptron neural network.
- The reported inference engine shows life-long lossless inference operation and (c). Reducing the ADC precision to 1bit causes a mere accuracy degradation of 1.87% from FP precision.
- 1-bit ADC shows less than 2% accuracy loss over 10 years.

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Benchmarking



Device Type	Fe-FinFET [9]	IWO-FeFET [14]	This Work
M3D Integrator	No	Yes	Yes
Cell Area (F ²)	15F ²	15F ²	16F ²
R _{on} (Ω)	100K	4M	100M
MW @10 years	1	0.2	1
Inference Accuracy Drop @10 years	~85%	85%	1.5%
Energy Efficiency (TOPS/W)	N/A	71.04	95.33 (Binary)

Acknowledgement

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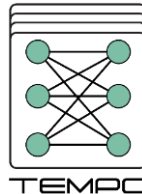


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Thank You

For your attention

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