

A 16-channel, 1-Second Latency Patient-Specific Seizure Onset and Termination Detection Processor with Dual Detector Architecture and Digital Hysteresis

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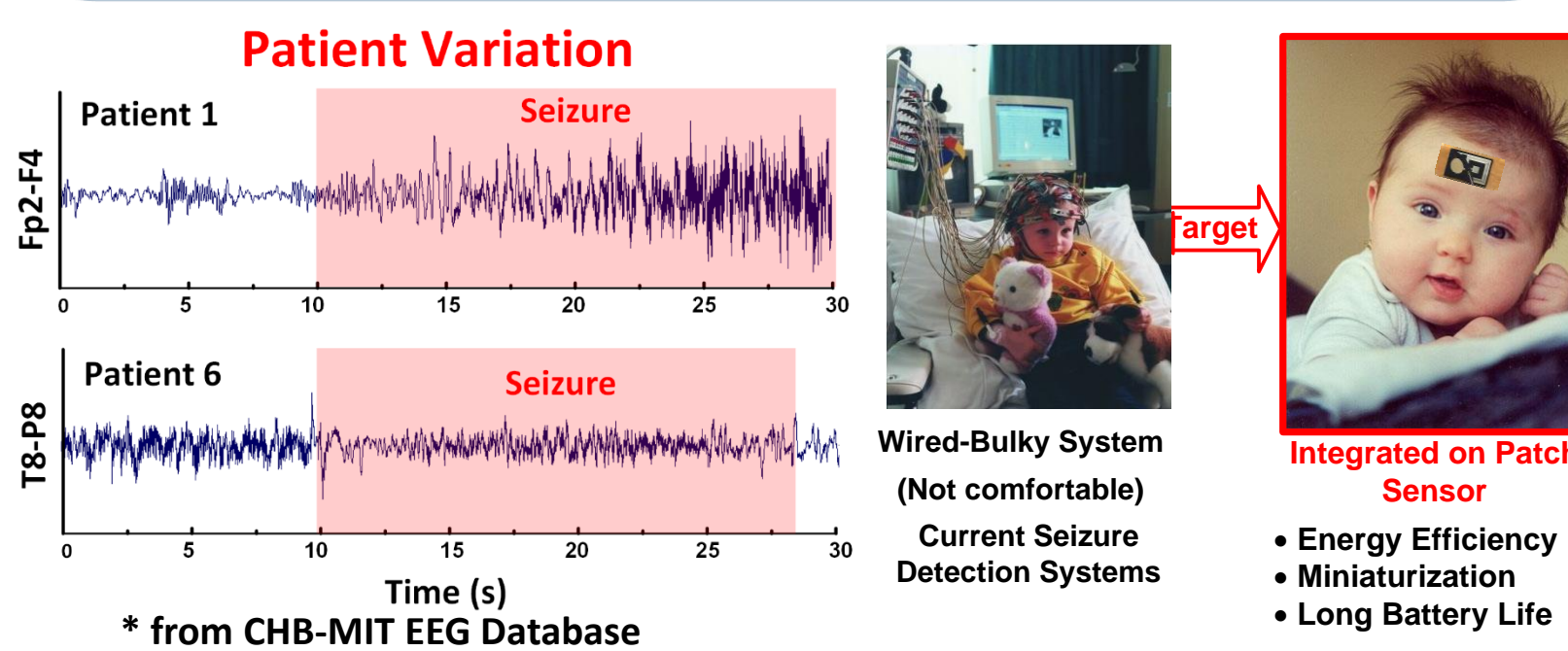
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I. Motivation

Some Facts About Epilepsy

- > **65 million** people affected worldwide^[1]
- > ~ **1.2 million** Patients Do NOT Benefit From Anti-Epileptic Drugs
- > Patient-to-Patient and Age-to-Age Variation is Huge
- > Min. **16-ch** EEG needed for Seizure Detection^[2]
- > Termination detection is needed for controlling electrical stimulation or drug release



II. Problems & Solutions

Issues with Existing Designs

- ⊗ Limited # of channels (≤ 8)
- ⊗ High power consumption^{[7]-[8]}
- ⊗ High resource utilization^[5]
- ⊗ High system latency (2s)^{[4]-[5]}
- ⊗ High detection Delay (13.8s)^[6]
- ⊗ Low accuracy & high false alarms^[4]
- ⊗ No seizure termination detection

S1

S2

S3

S4

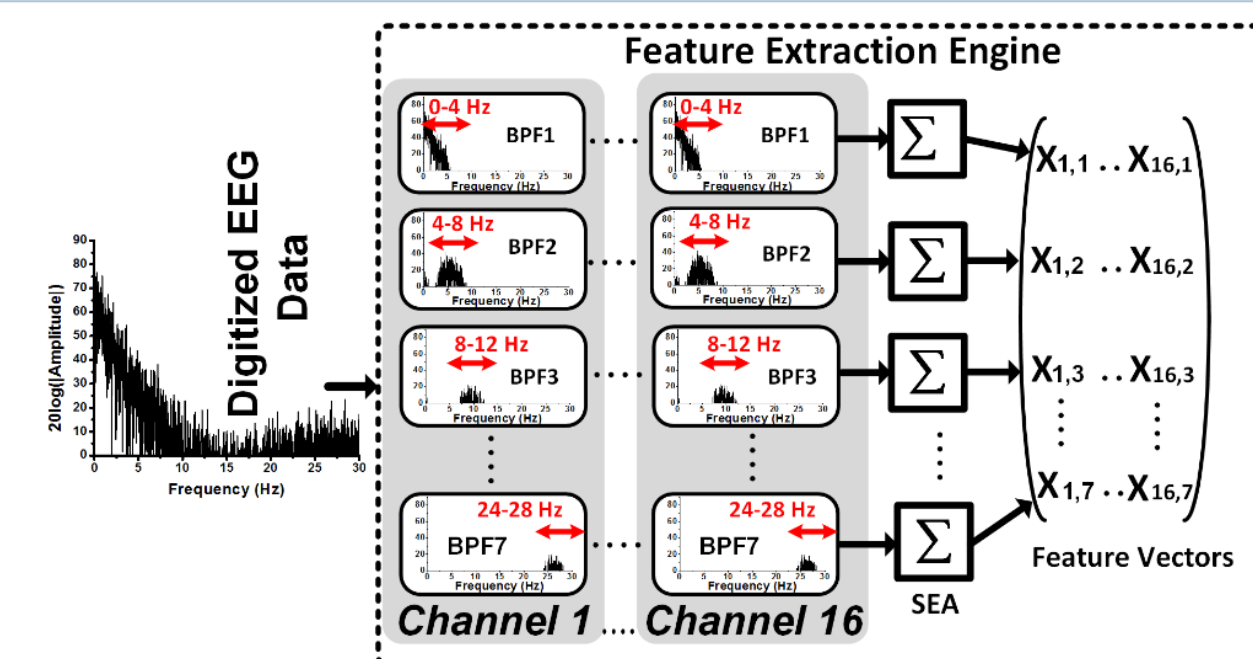
Solutions

- > S1. Frequency-Time Division Multiplex BPF
- > S2. Feature Extraction with Overlapped-Sliding Window
- > S3. D²A-LSVM architecture + Digital Hysteresis
- > S4. Weight-And-Average (WAA) Algorithm

III. Feature Extraction (FE) Engine

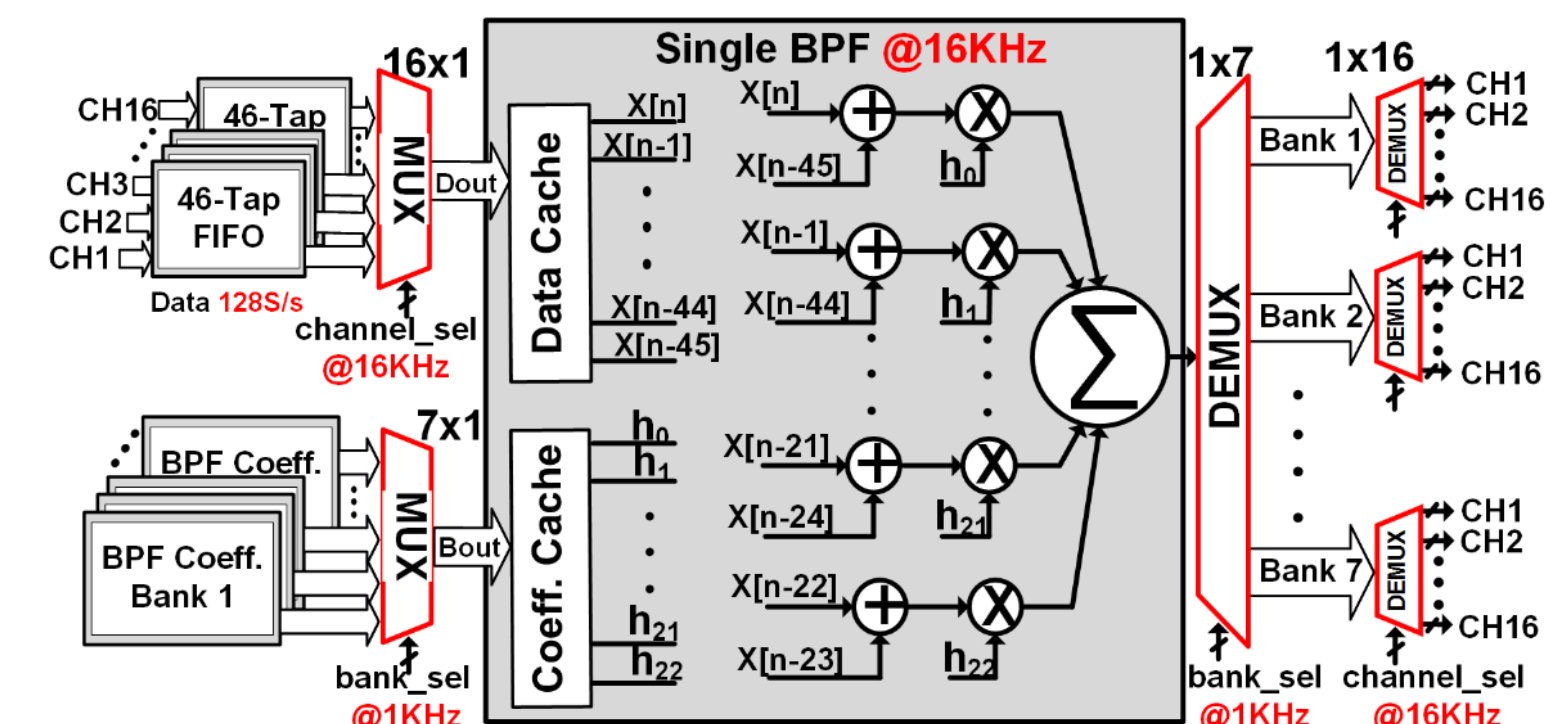
Feature Vector = Spatial + Temporal + Spectral

- > 7 BPF*4Hz = 28Hz + Spectral Energy Accumulator(SEA)



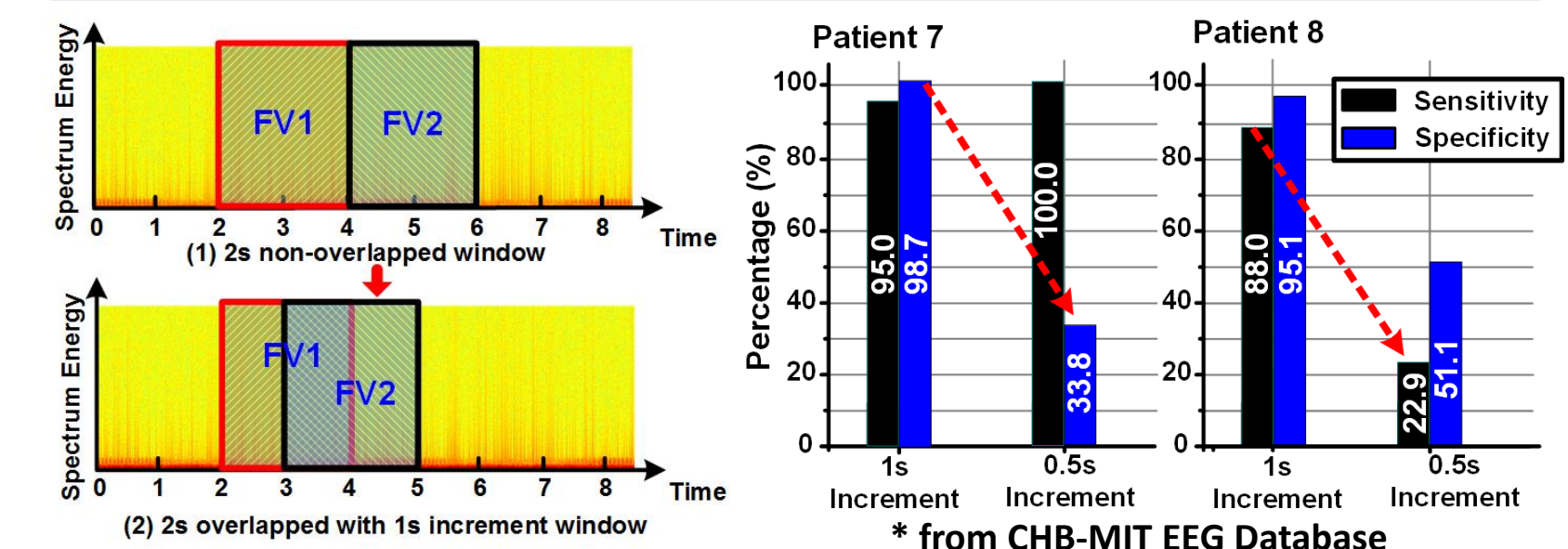
Frequency-Time Division Multiplex BPF Architecture

- > 1 BPF for 112 Banks in 16 Channels
- > Gates: **40.5%** ↓ compared to TDM-BPF^[5];
- > Power: **38.4%** ↓ compared to TDM-BPF^[5];



2s Overlapped-Sliding Window with 1s Increment

- > 50% Improvement in System Latency (**2s**→**1s**)
- > No Compromise in Classification Accuracy



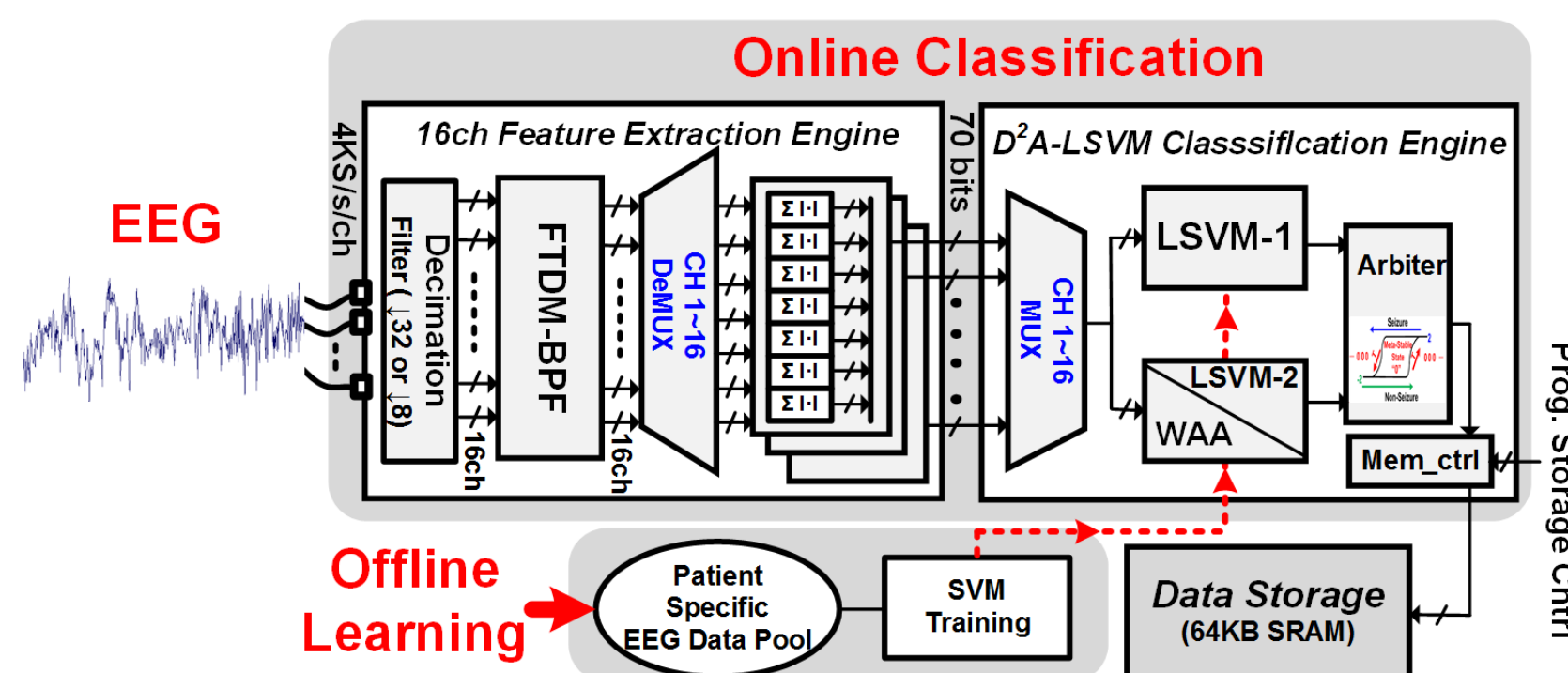
IV. D²A Classification Engine

Conventional SVM-Based Seizure Detector

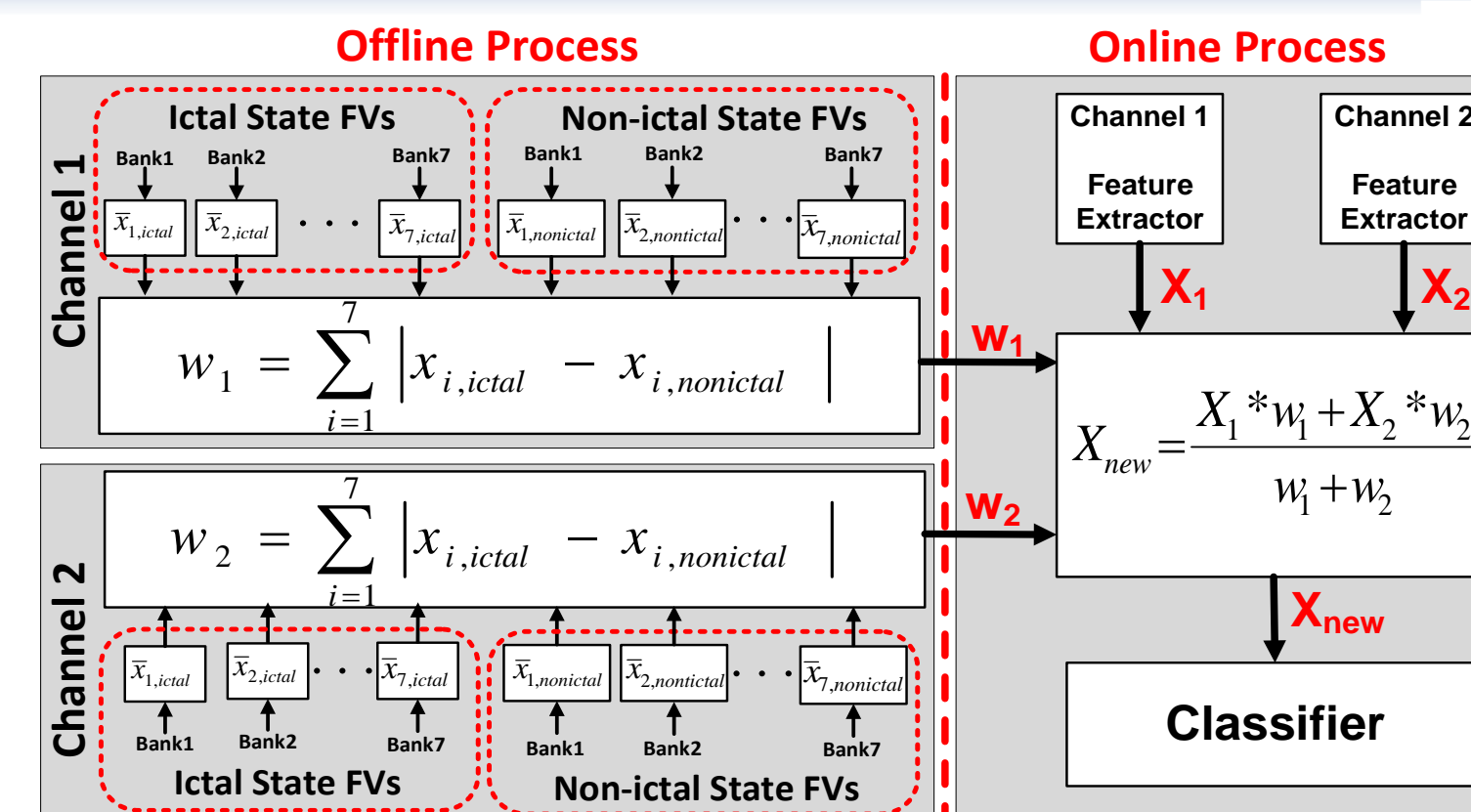
- > Single LSVM detector^[9] → ⊗ Low Accuracy
- > Single Non-Linear SVM^[11] → ⊗ 100's of SVs

Proposed D²A-LSVM Architecture (Two LSVMs)

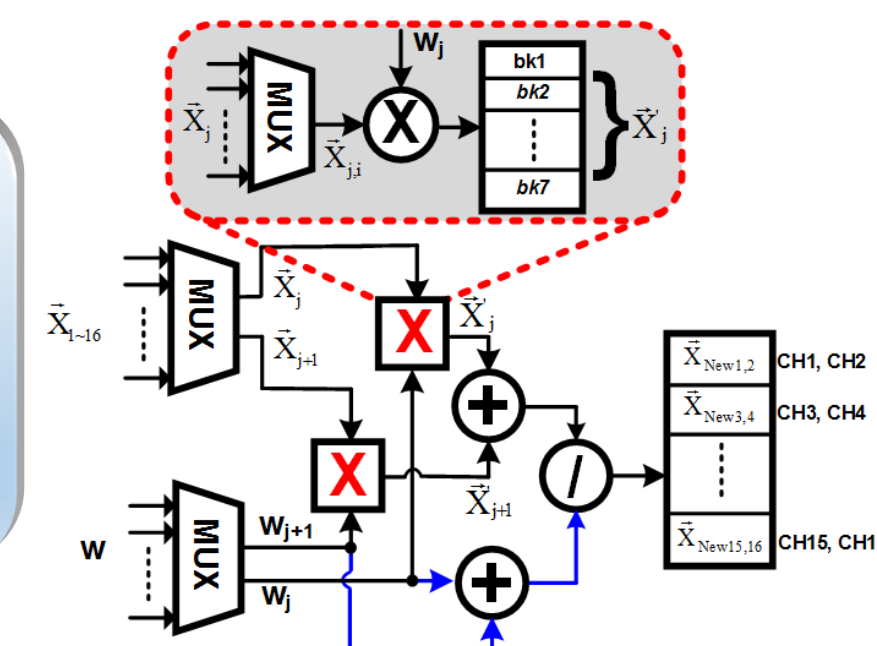
- > Detector-1 → Optimized for **Sensitivity**
- > Detector-2 → Optimized for **Specificity**



a) Weight-And-Average Algorithm



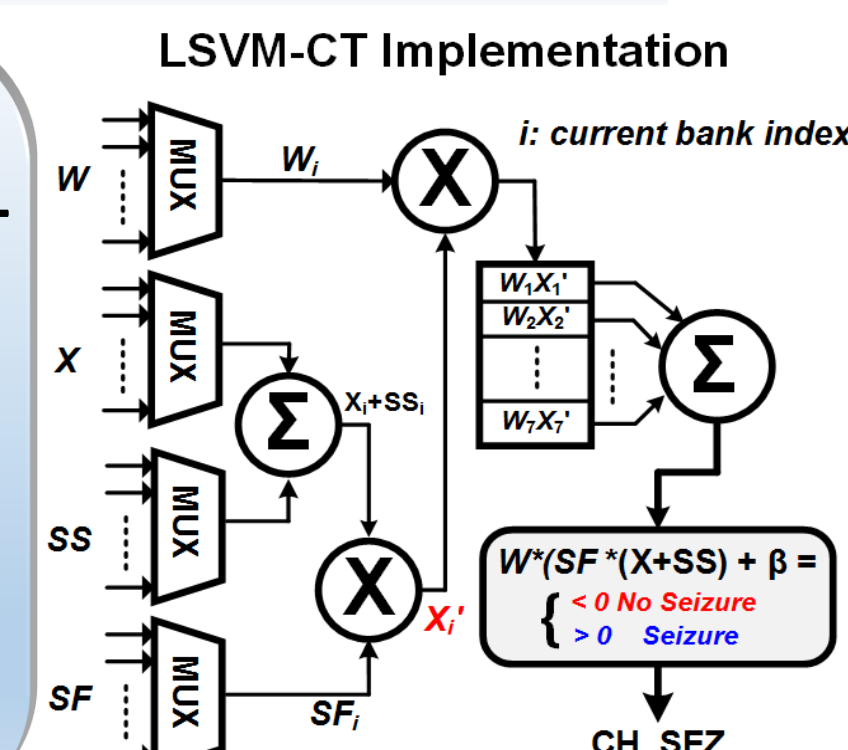
- > WAA Implementation using TDM
- > Gates: **56.0%** ↓
- > Power: **27.5%** ↓



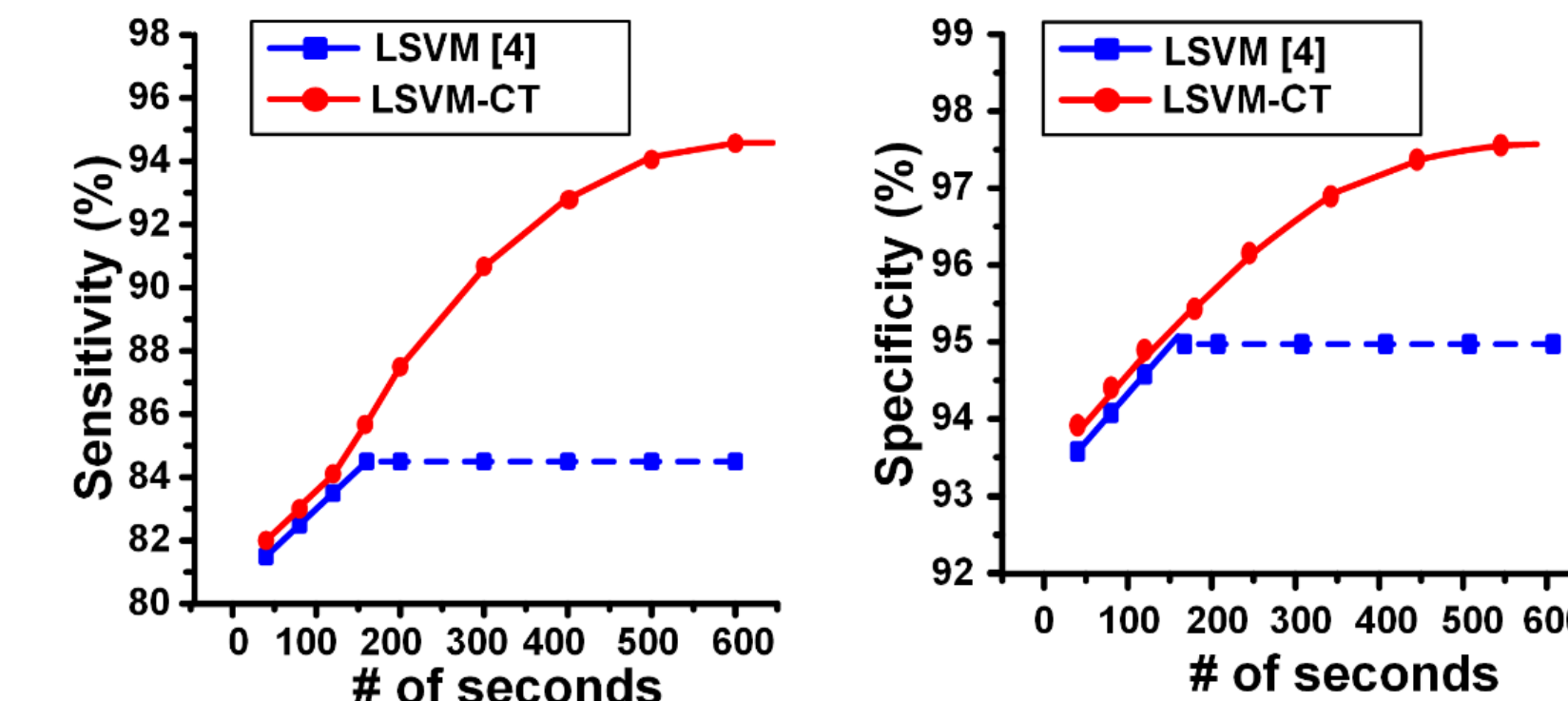
IV. D²A Classification Engine (contd.)

b) Linear SVM with Condensed Training (LSVM-CT)

- > More Tolerance Margin
- > Data Compression for CT
- > Sensitivity: **84.4%** → **93.5%**
- > Specificity: **95.0%** → **97.6%**
- (Results based on Single LSVM-CT)

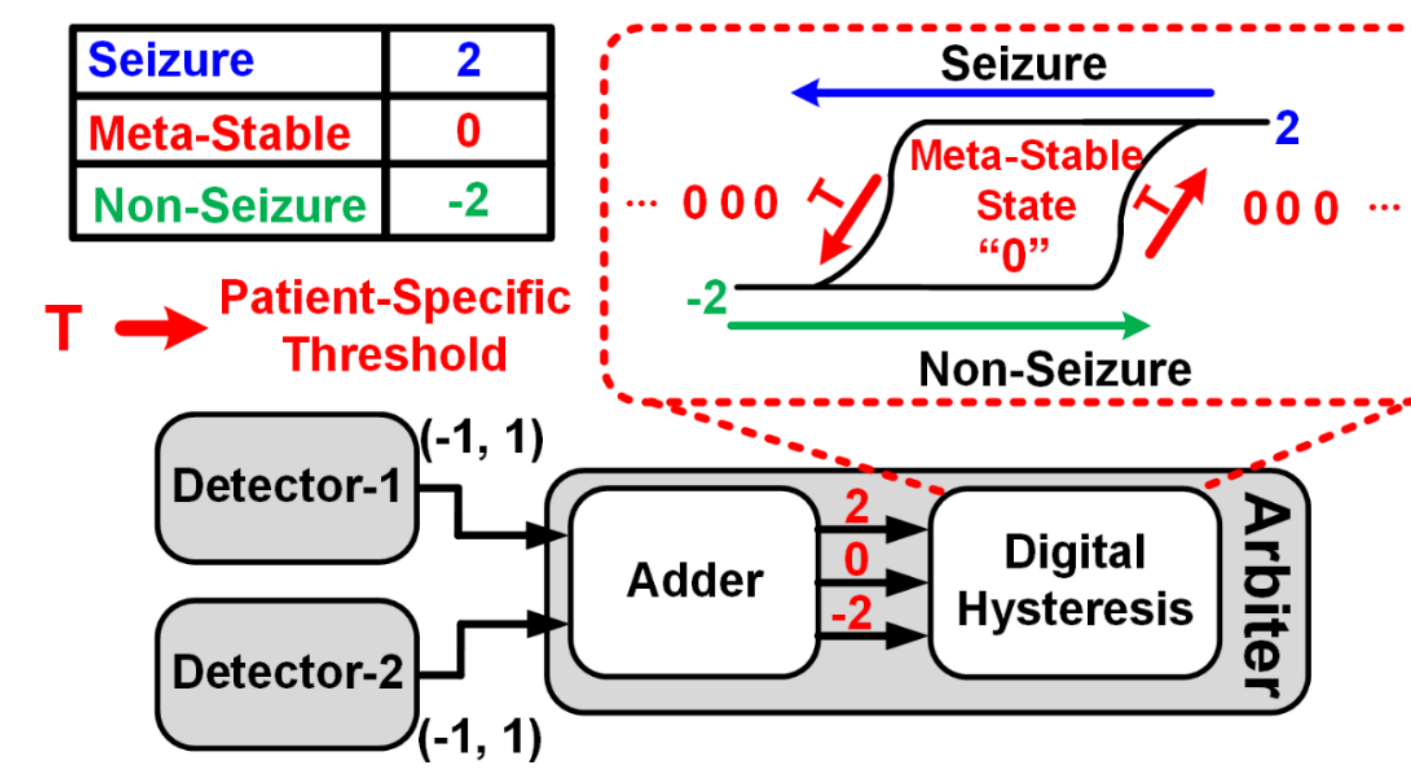


(a) LSVM-CT vs LSVM



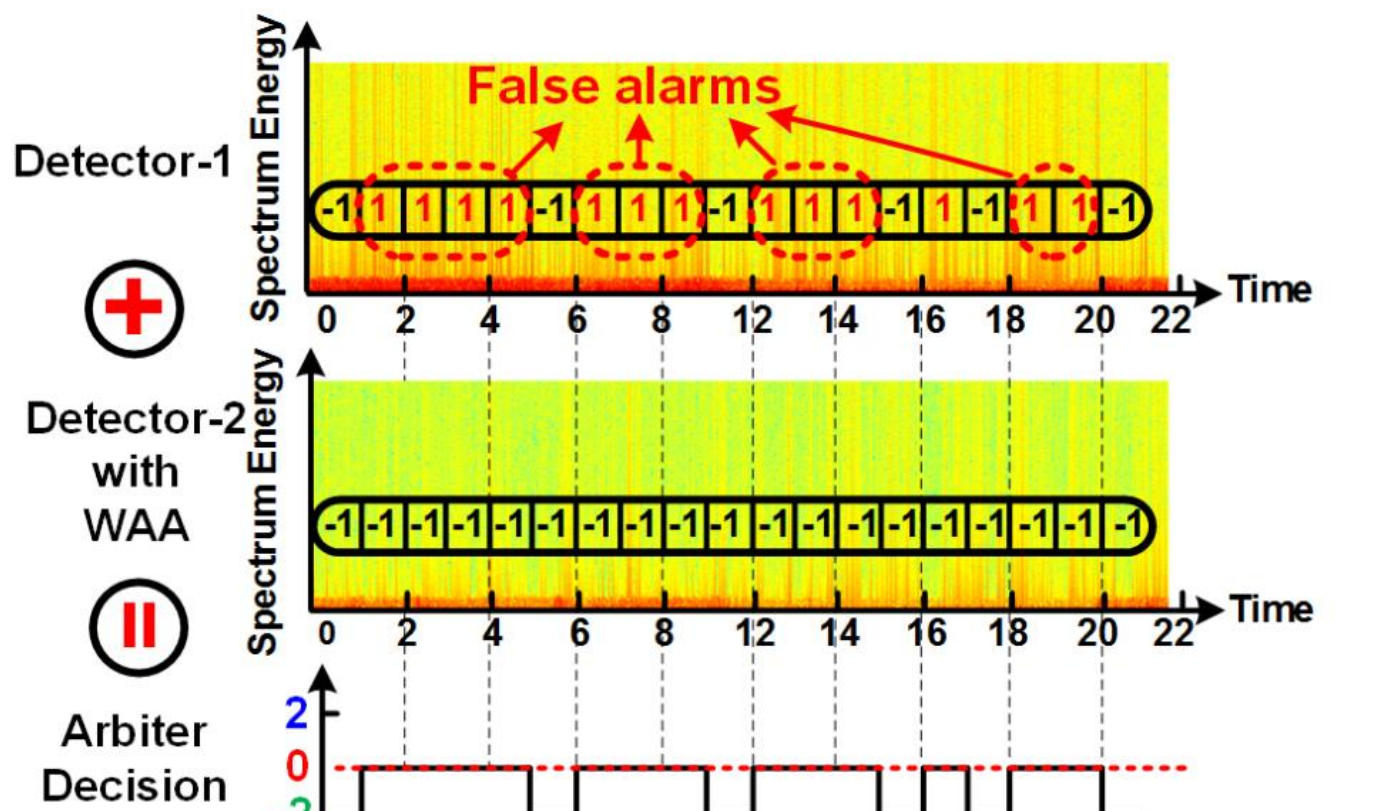
c) Arbiter with Digital Hysteresis

- > Combines both detectors' advantages
- > Improves Sensitivity and Specificity simultaneously
- > Decreases overall false alarms

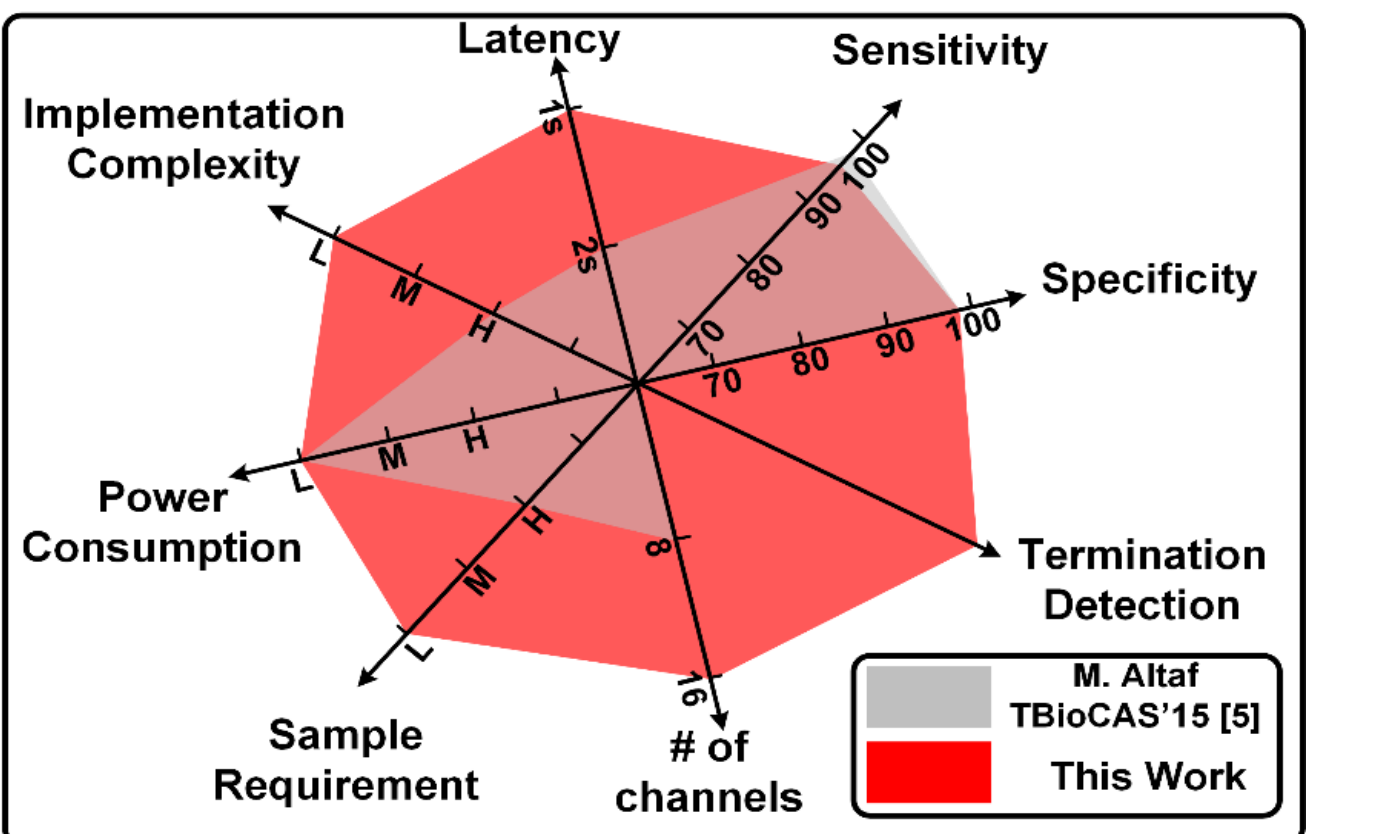


V. Implementation Results

False alarm elimination with digital hysteresis



Seizure detection SoC Comparison



Parameter	JSSC'13 [4]	ISSCC'13 [5]	TBioCAS'13 [6]	JSSC'14 [7]	JSSC'13 [8]	This Work
Patient Specific	Linear SVM	Gaussian SVM	HFD	LLS	SVM	D ² A-LSVM
Seizure Termi. Detection	X	X	X	X	X	O
Sensitivity	84.4%	>96%	100%	92%	100% (by event)	95.7%
Specificity	96%	>99%	100%	--	0.75 FA/H (0.27 FA/H)	98.0%
Onset Delay	--	--	13.8s	--	15s	3.4s
System Latency	2s	2s	--	0.8s	--	1s
No. of Channels	8	8	1	8	1	16
On Chip Storage	O	O	X	X	X	O
Energy Efficiency	1.49uJ/Class	1.31uJ/Class	44.85mW	77.9uJ/Class	273uJ/Class	1.85uJ/Class
Technology	0.18um	0.18um	0.18um	0.18um	0.13um	0.18um

References

- [1] A. Shoeb, Ph.D. Thesis, MIT.
- [2] American Clinical Neurophysiology Society
- [3] M. Altaf et al, ISSCC Dig. Tech. Papers, 2015.
- [4] J. Yoo et al, IEEE JSSC, 2013.
- [5] M. Altaf et al., in ISSCC Dig. Tech. Papers, 2013.
- [6] M. Mirzaei et al., IEEE Trans. Biomed. Circ. Syst, 2013.
- [7] W. Chen et al., IEEE JSSC, 2014.
- [8] K. Lee et al., IEEE JSSC, 2013.