# Energy-Efficient Discrete Signal Processing with Field Programmable Analog Arrays (FPAAs)

Yu Bai and Mingjie Lin



**Stands For Opportunity** 

# Why Convolution?

- Large-Scale Convolution is the Computing Engine of many CV, AI, and DSP application
  - Image Classification, Edge Detection, ...
  - Deep Learning: Convolutional 3D-Network, …
  - Video Object Recognition, ...
- But, Computationally Intensive
  - Energy Consumption High
  - Fault Tolerance Low
  - Time Complexity O(n<sup>2</sup>) or O(nlg(n)) even with complicated FFT-based method

# **Analog Computing**

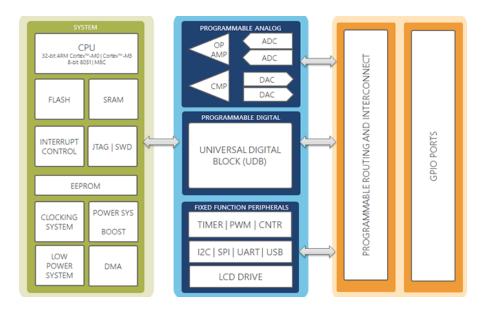
- Digital Computing vs. Analog Computing
  - Energy Efficiency
  - Performance
  - Hardware Effectiveness



- Analog Computing is HARD!
  - Flexible and Effective Prototyping Platform is Lacking
  - Resource Constraints
  - Noise Sensitivity
  - No Unified Computing Framework for Complicated Task

## What about Reconfigurable Analog Computing?

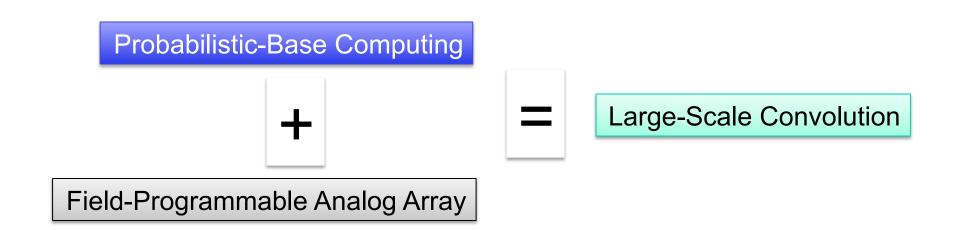
Limited Resource Type: No Universal Logic Gate Set, No LUT ....



No Straightforward Design Framework: No Logic Synthesis + Technology Packing ....

Very Sensitive to Noise and Design Parameters: The General Nature of Analog Computing ...

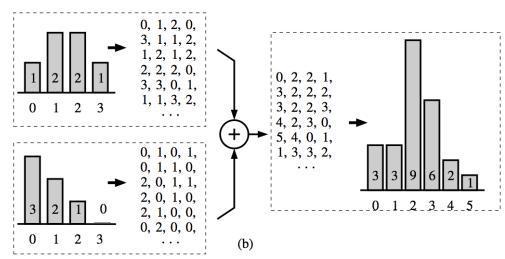
## Our Idea



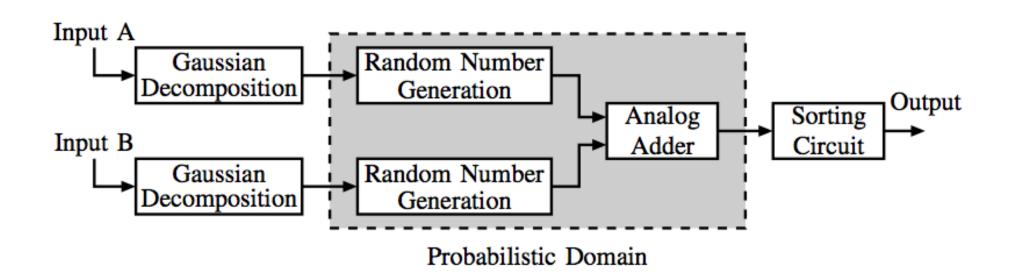
- Prototyping
- Resource Constraints
- Noise Sensitivity
- No Unified Computing Framework

#### **Probabilistic Method for Convolution**

Probabilistic Convolution Theorem<sub>z</sub>  $f_Z(z) = f_{X+Y}(z) = P(Z = z) = \sum f_X(x)f_Y(z-x)$ x=0z|n|11 x[n] $y[n]_{2}$ 2 2 \* 3 n0 1 2 3 0 1 2 0 1 2 3 4 5 3 (a)

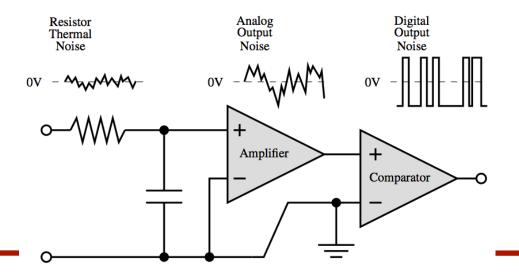


## **Overall Algorithm Flow**

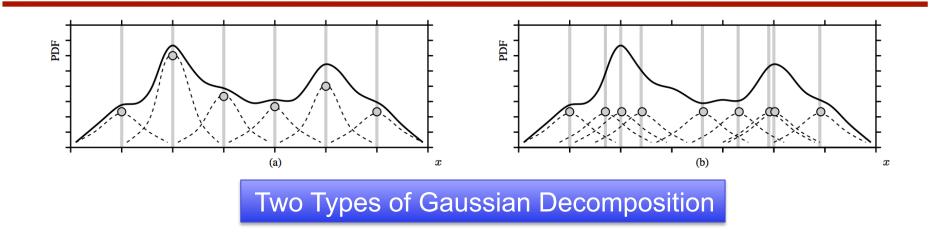


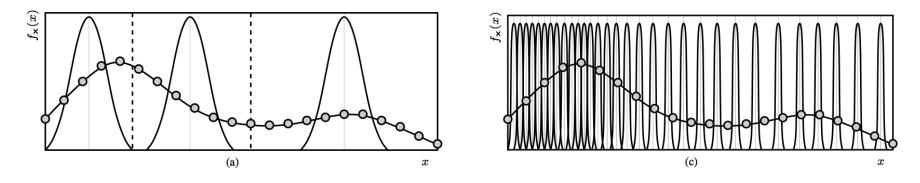
## Input $\rightarrow$ PDF $\rightarrow$ Random Samples

- Direct Implementation too costly
- Objective: Given any form of PDF, efficiently generating a large ensemble of random samples
- Method:
  - Gaussian noise source + Gaussian Decomposition
- Gaussian Noise Random Sample Generator



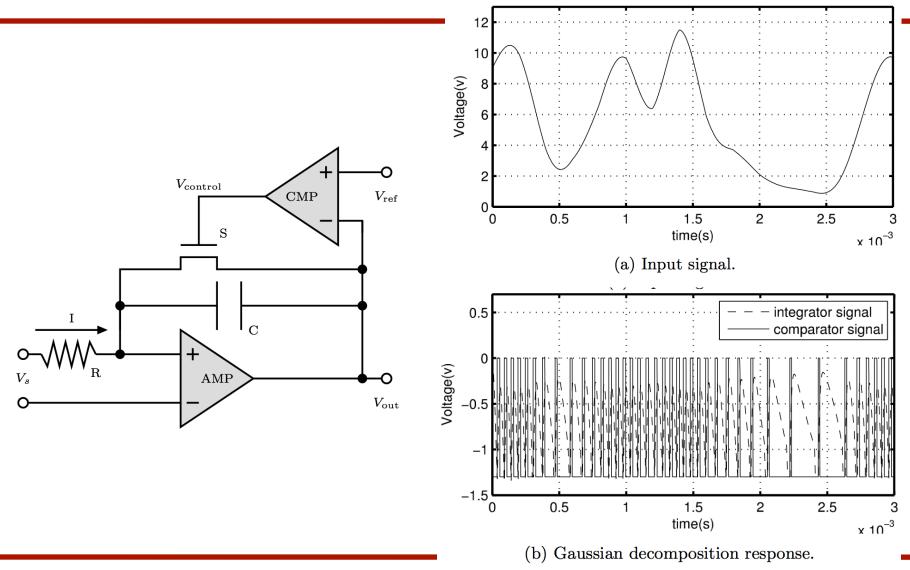
### **Gaussian Decomposition: Concept**



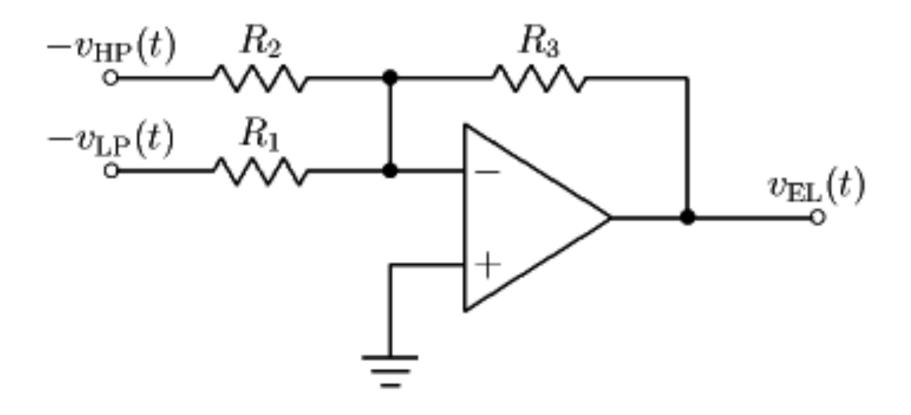


Tunable Accuracy: Number of Gaussian Components

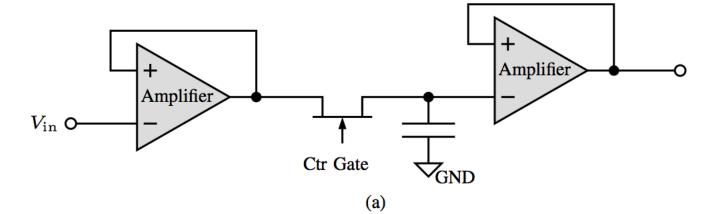
## Gaussian Decomposition: Analog Way

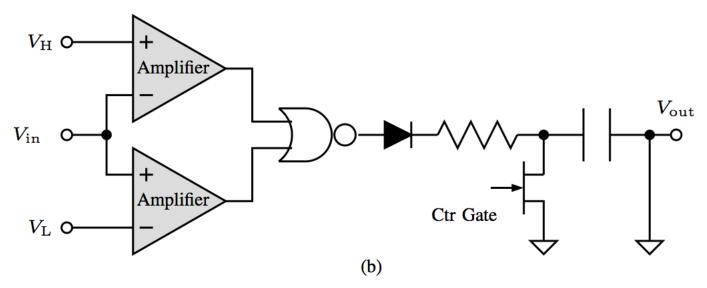


## Analog Adder

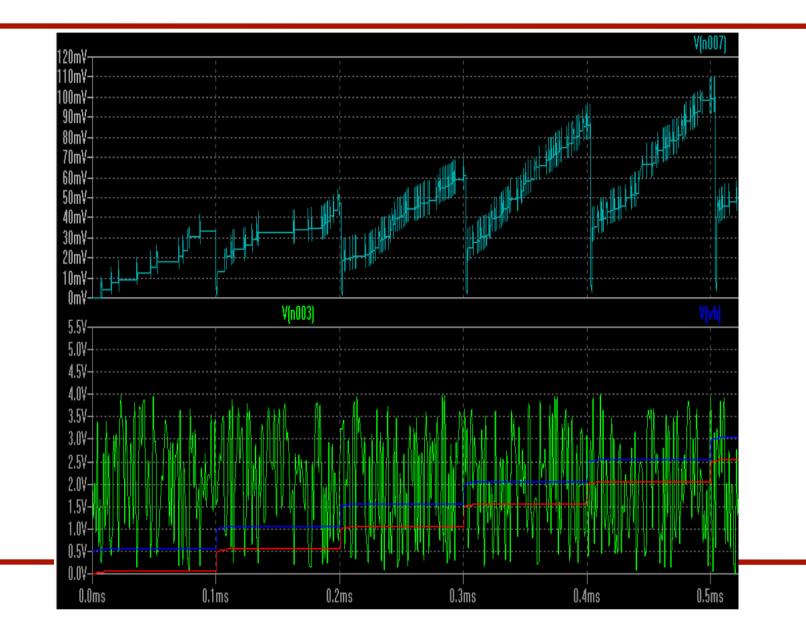


### Random Samples $\rightarrow$ PDF: Analog Way



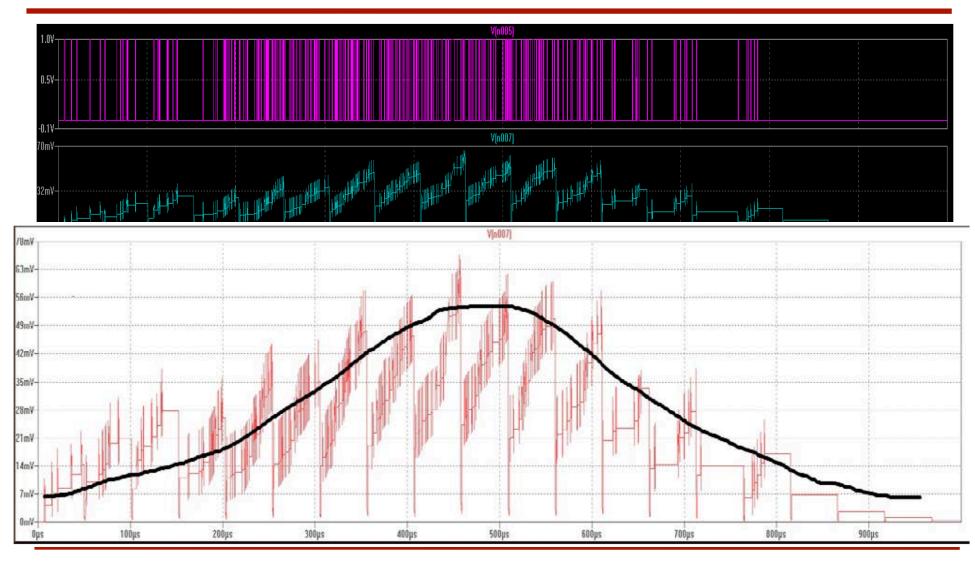


#### Random Samples $\rightarrow$ PDF: Results

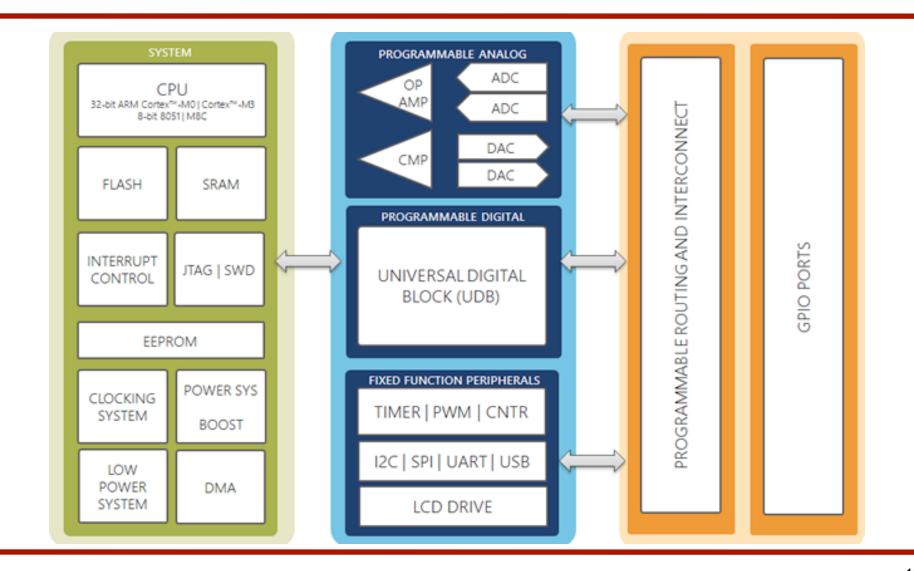


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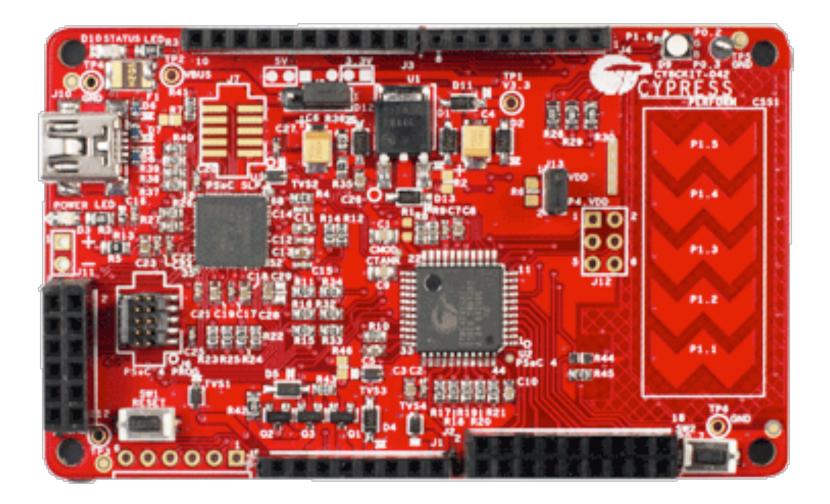
## Random Samples $\rightarrow$ PDF: Results



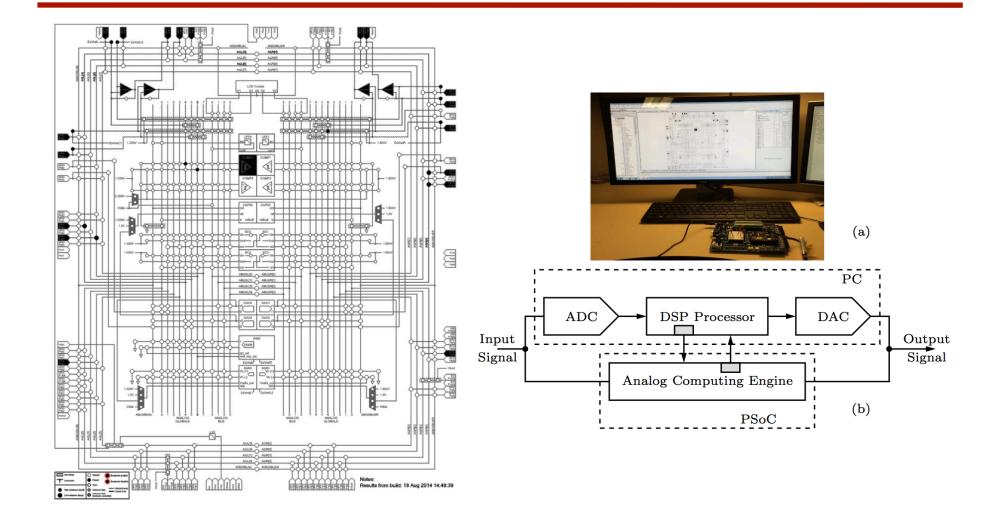
#### **PSoC Platform for Implementation**



## **PSoC Board**



#### **PSoC System Implementation**

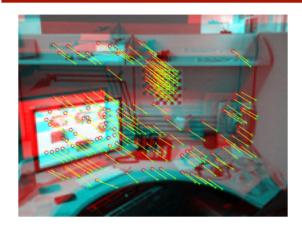


## Hardware Usage and Energy Consumption

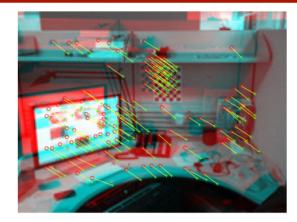
		FPGA-Based			PSoC-Based		
		64	1024	4096	64	1024	4096
Area	Slice LUT	950	3308	13232			
	Slice Registers	426	1436	5744			
	Slice	310	1036	4144			
	Opamp				4	6	8
	comparator				1	1	1
	capacitor				4	8	12
	resistor				5	9	14
	diode transistor				4	7	13
Power	Dynamic Power	107.82	137.87	186.28	46.66	55.29	64.88
	(mW)						
	Energy Consumption	1196.80	1098.82	830.81	427.24	392.43	153.86
	(nJ)						

~1/3 of Dynamic Power, ~1/6 of Total Energy Consumption at 99% accuracy

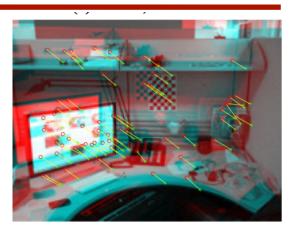
## App1: Finding Corresponding Features



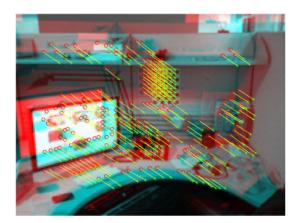
(a)  $\sigma=1\%$ , P=185.



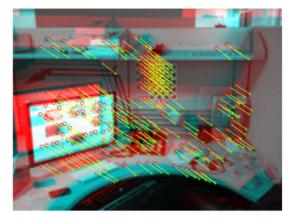
(c)  $\sigma$ =5%, P = 114.



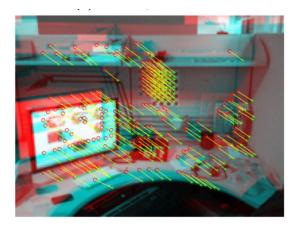
(e)  $\sigma$ =10%, P = 60.



(b)  $\sigma$ =1%, P = 234.



(d)  $\sigma$ =5%, P = 230.



(f)  $\sigma$ =10%, P = 220.

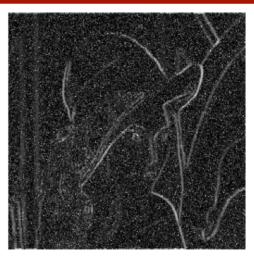
#### App2: Convolution-Based Edge Detection



(a)  $\sigma = 1\%$ .



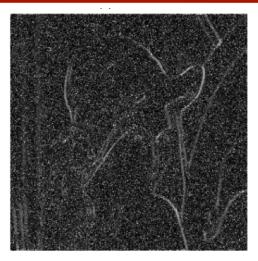




(c) *σ*=5%.





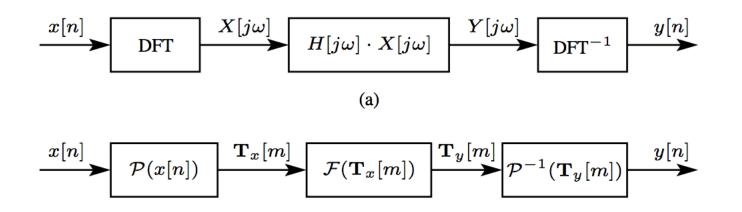


(e) *σ*=10%.



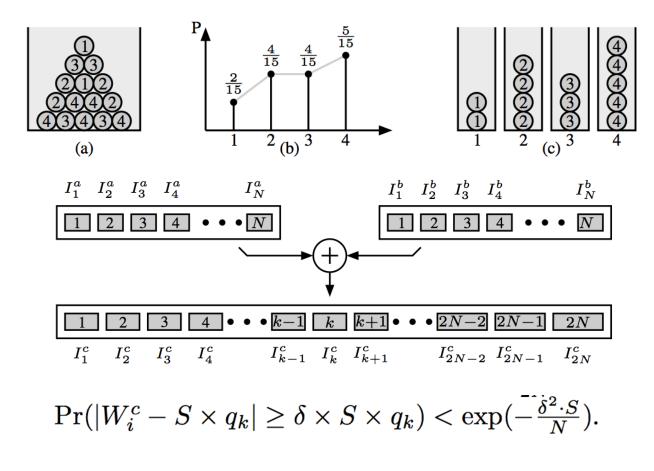


## **Question 1: How Versatile?**



- Computing Framework: Probabilistic Domain Transform
- Perception-Based Computing Tasks
  - Convolution-Like Applications: FIR, Correlation, ...
  - Associative Computing, …

#### **Question 2: How Accurate?**



Chernoff Bound → # Random Samples vs. Accuracy

## Thanks! Questions?