

Event-Based Re-configurable Hierarchical Processors for Smart Image Sensors

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Motivation and Challenges

1. Continuous Increase in image size
2. Sequential operation
3. High power consumption
4. Application specific image sensor

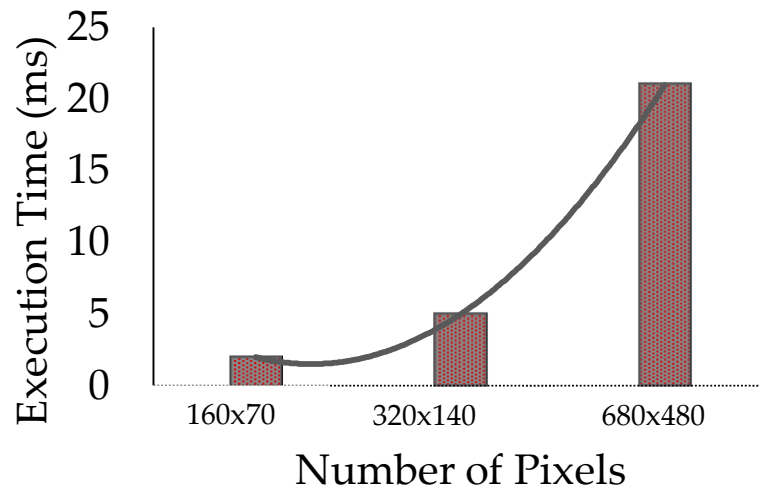


Figure 2: Increase of execution time with the increase of pixel counts

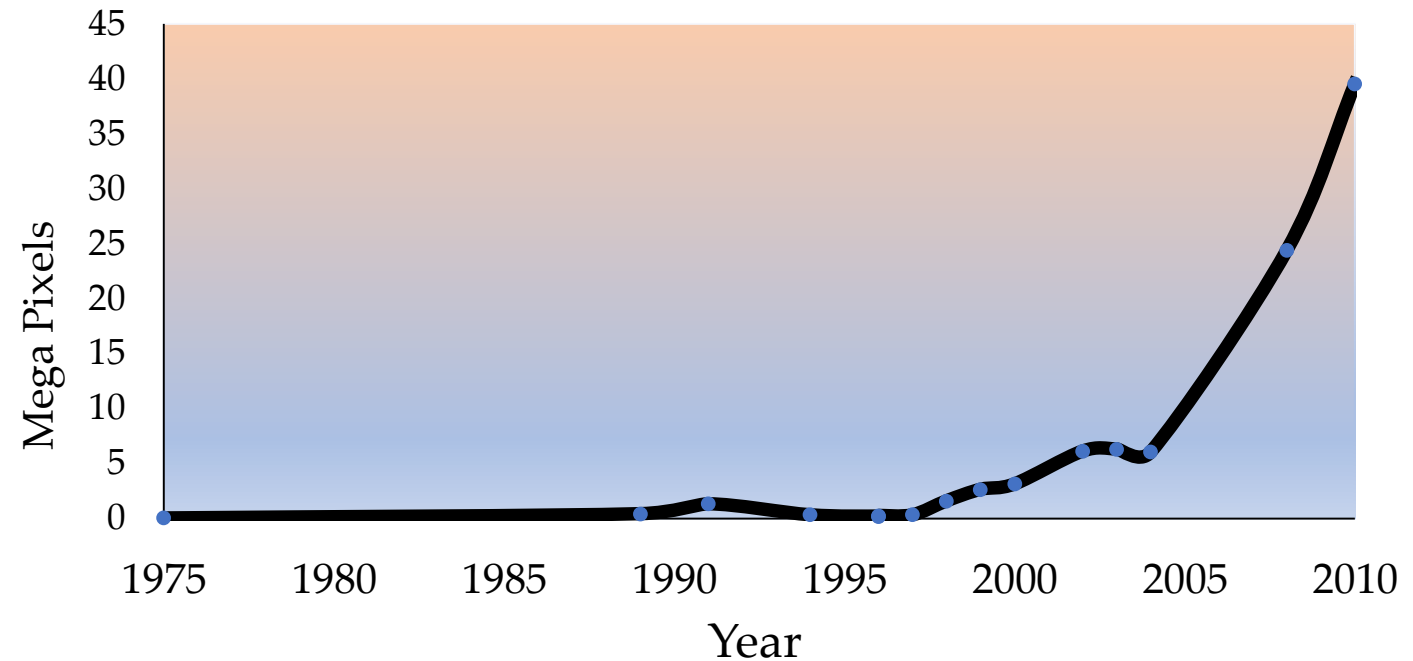


Figure 1: Pixel Size Increase over the years in Digital Camera*

Viabile Solutions

- Apply event-based processing instead of frame-based processing.
- Bringing the computational units close of the image sensor.
- Break sequential nature of image processing and introduce massive parallelism.
- Bio-Inspired computing promises to reduce power consumption.
- Hardware should be reconfigurable in ASIC paradigm.

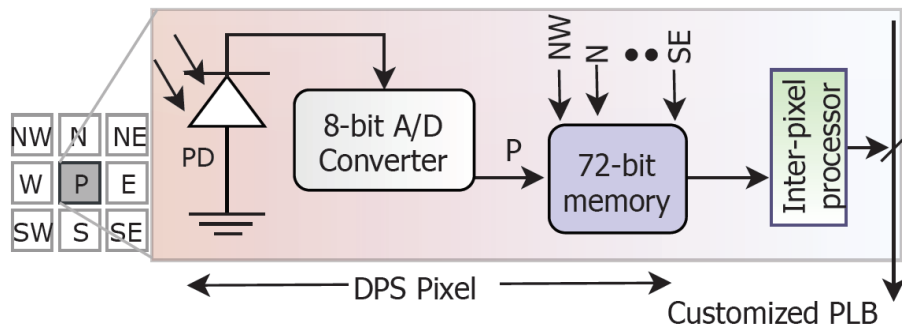


Figure 3: Pixel Circuit with computational unit

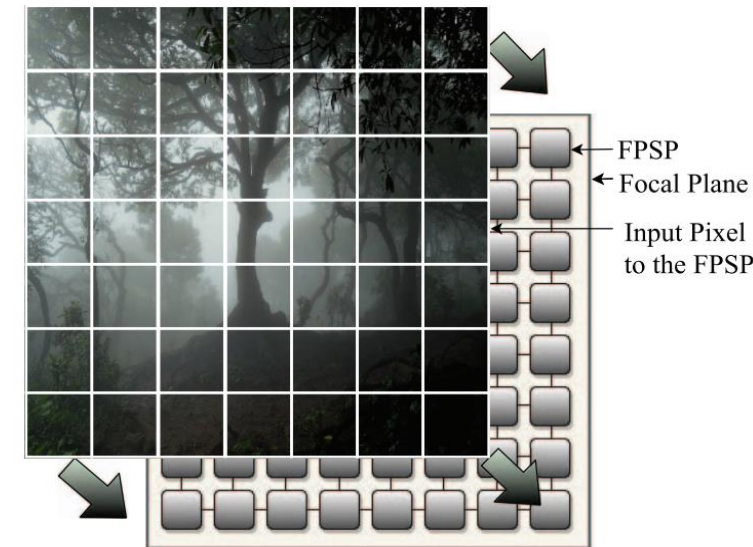


Figure 4: Pixel-Parallel Focal Plane Image sensor where computational units are brought close to the image sensor*

Why Event-Based Processing is inevitable?

Frame Based Processing

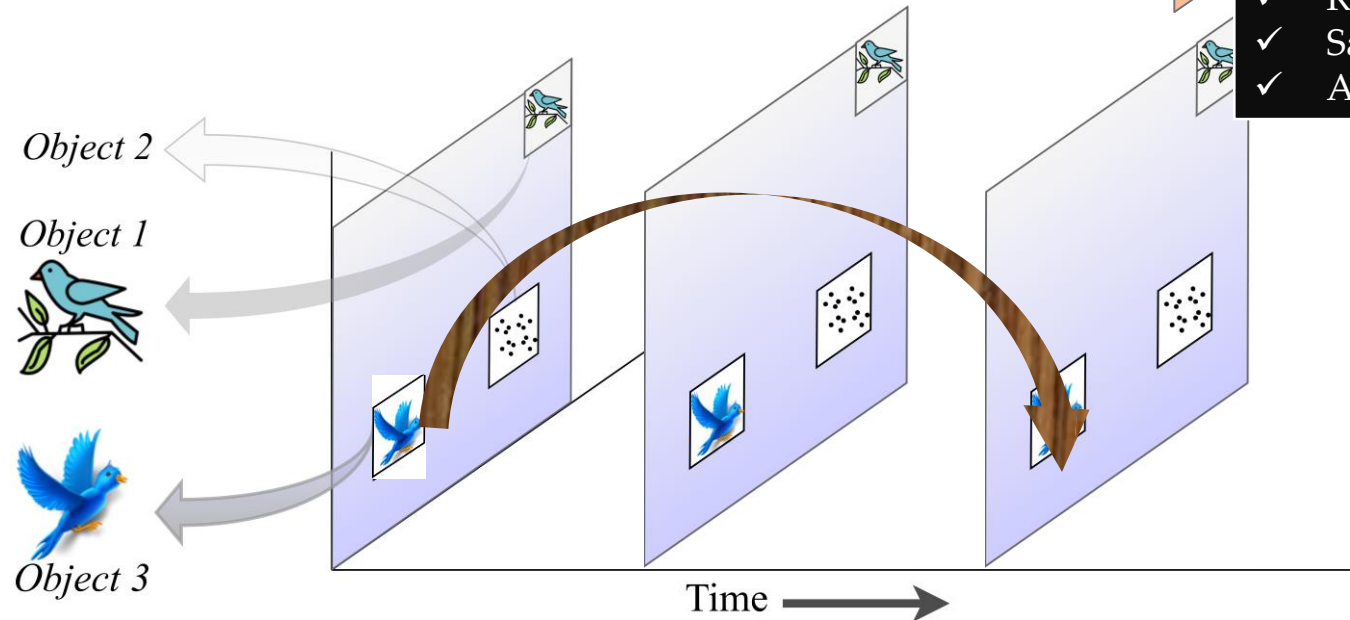
Every Pixel is computed sequentially

Event-Based Processing

Pixels with relevant information are computed in a sequential or parallel process

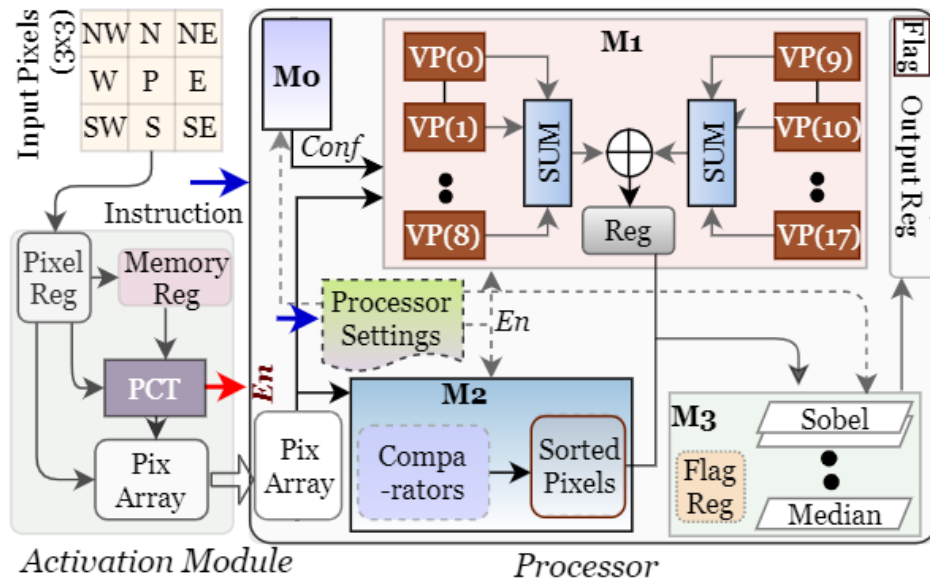
- ✓ Gleans effective information
- ✓ Reduces data volume
- ✓ Saves computational power
- ✓ Accelerate the processing

Convention
Event-
Based
Processing

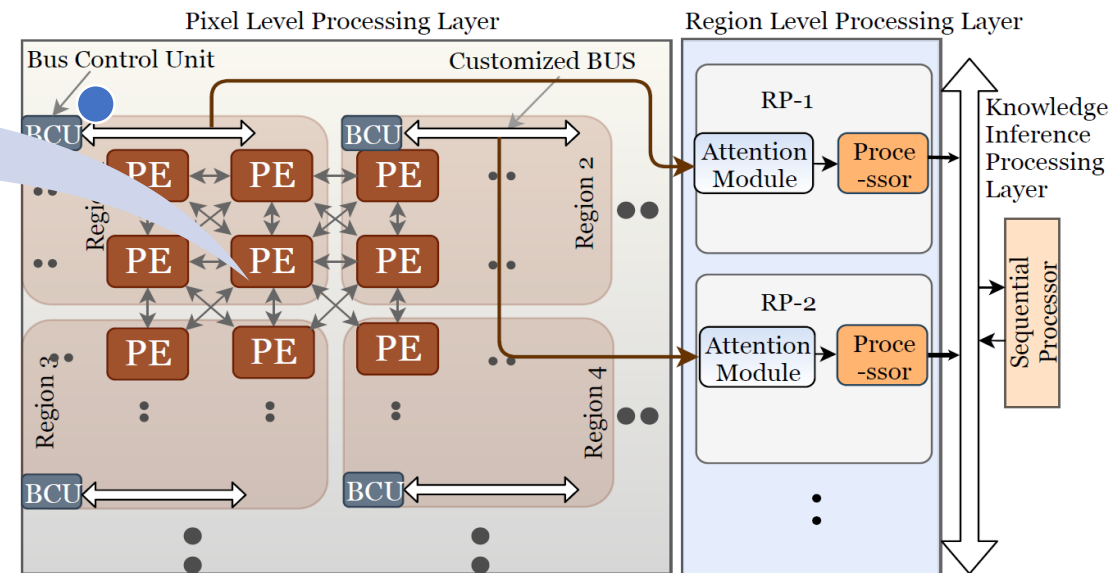
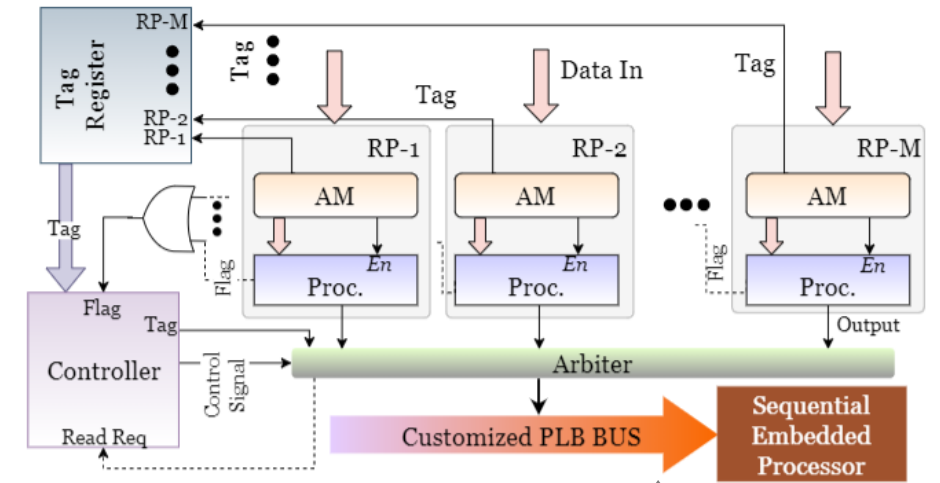


Proposed Architecture

1. Pixel-Level Processing Layer
2. Region-Level Processing Layer
3. Knowledge Inference Processing Layer

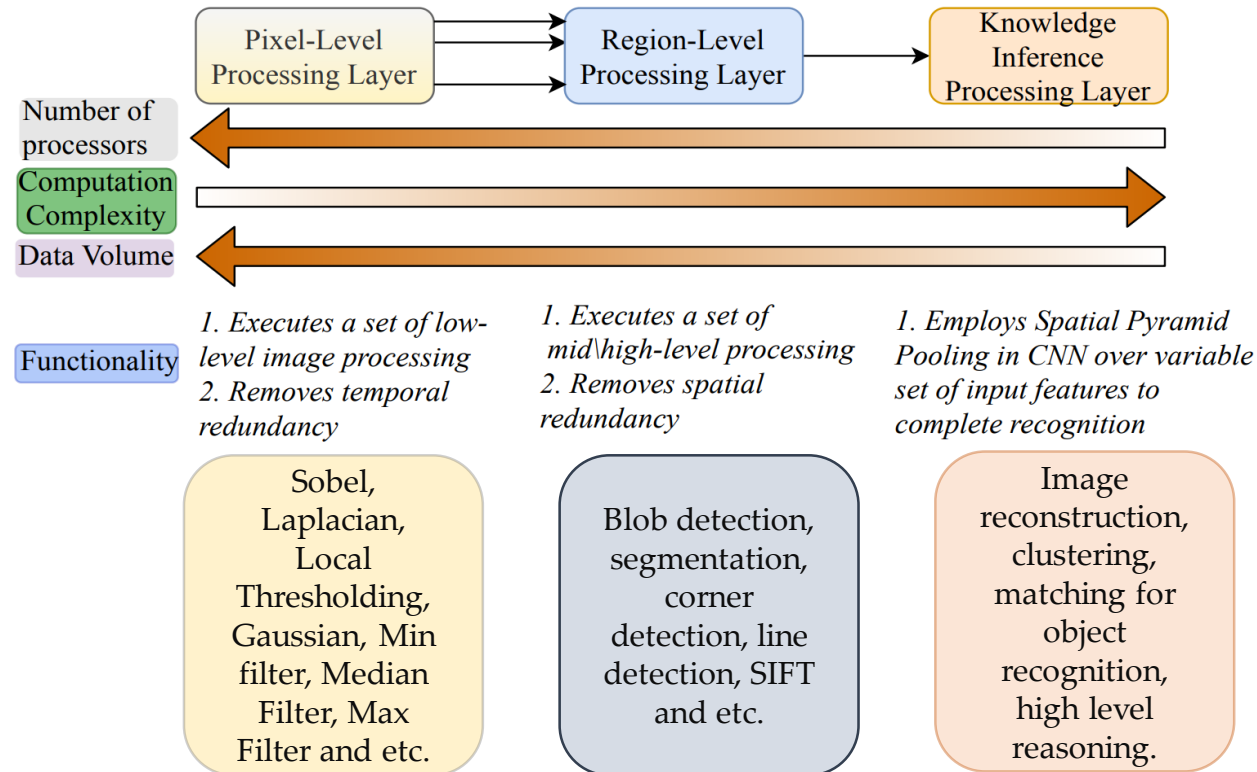
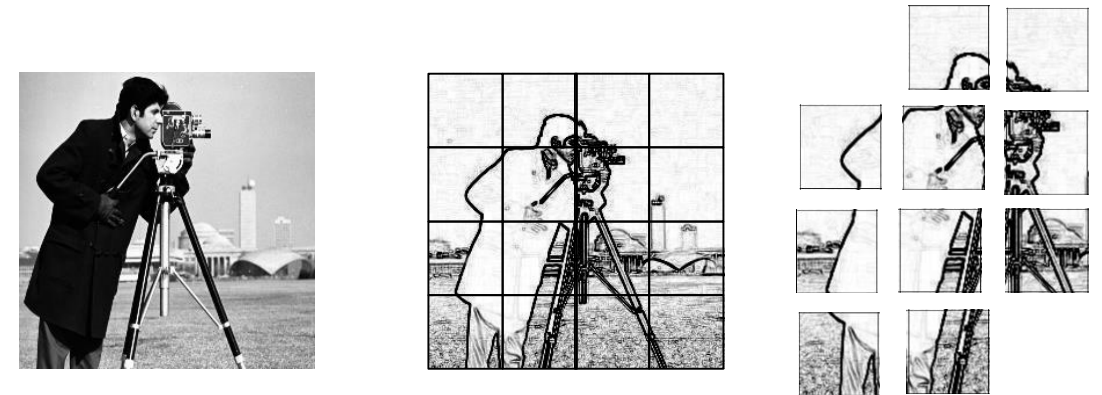
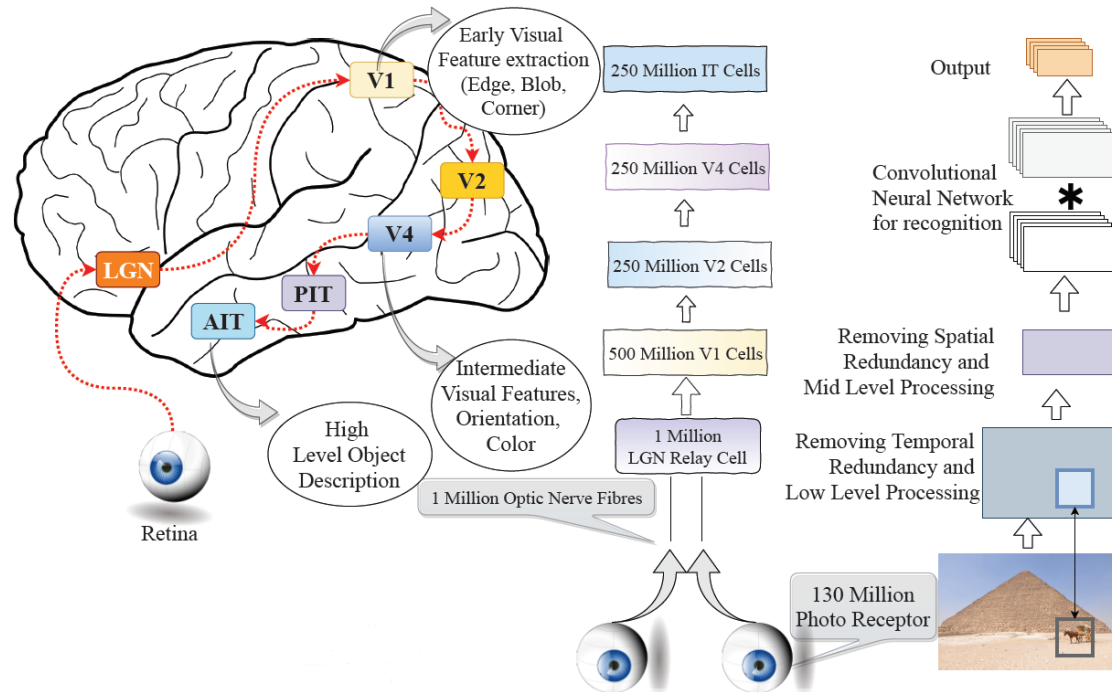


Single pixel circuit



Hierarchical Processing

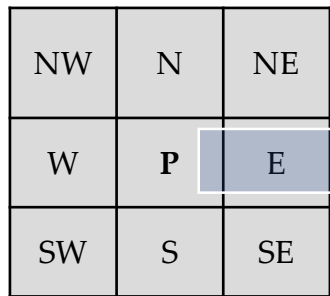
Biological Vision System is emulated in the proposed model



Pixel-Level Processing Layer

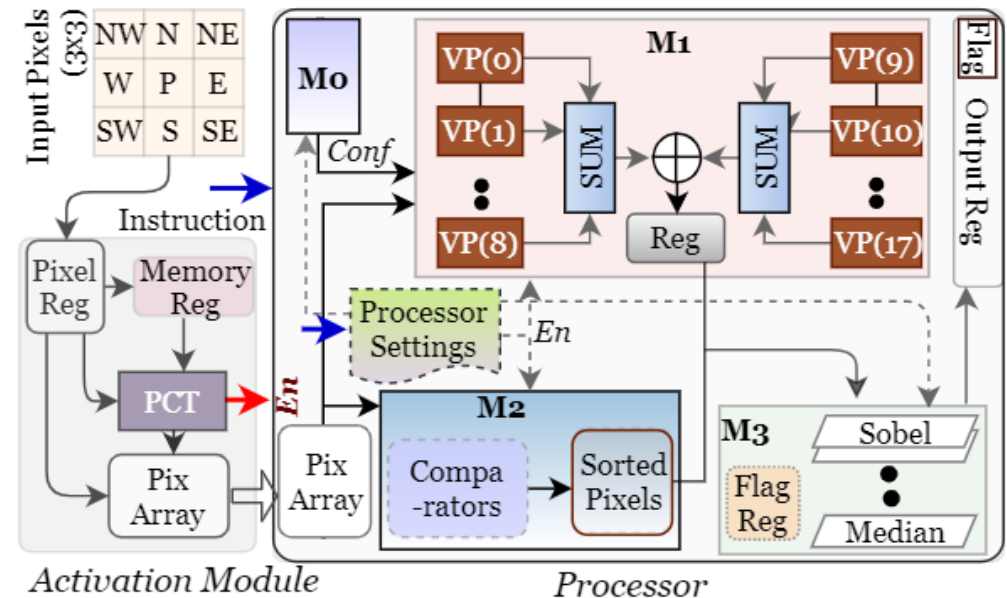
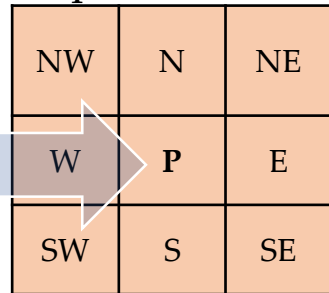
1. **Activation Module:** Predictive Coding in time
2. **Processor:** Reconfigurable to a set of Low-Level Image processing applications
 - M1 and M2 module executes the common section of the applications
 - M0 provides the convolutional kernel to the M1
 - M3 executes the unique segments
 - Processor setting block sets the configuration

Inputs to the PE works with pixel P at time $t-1$



PCT

Inputs to the PE works with pixel P at time t



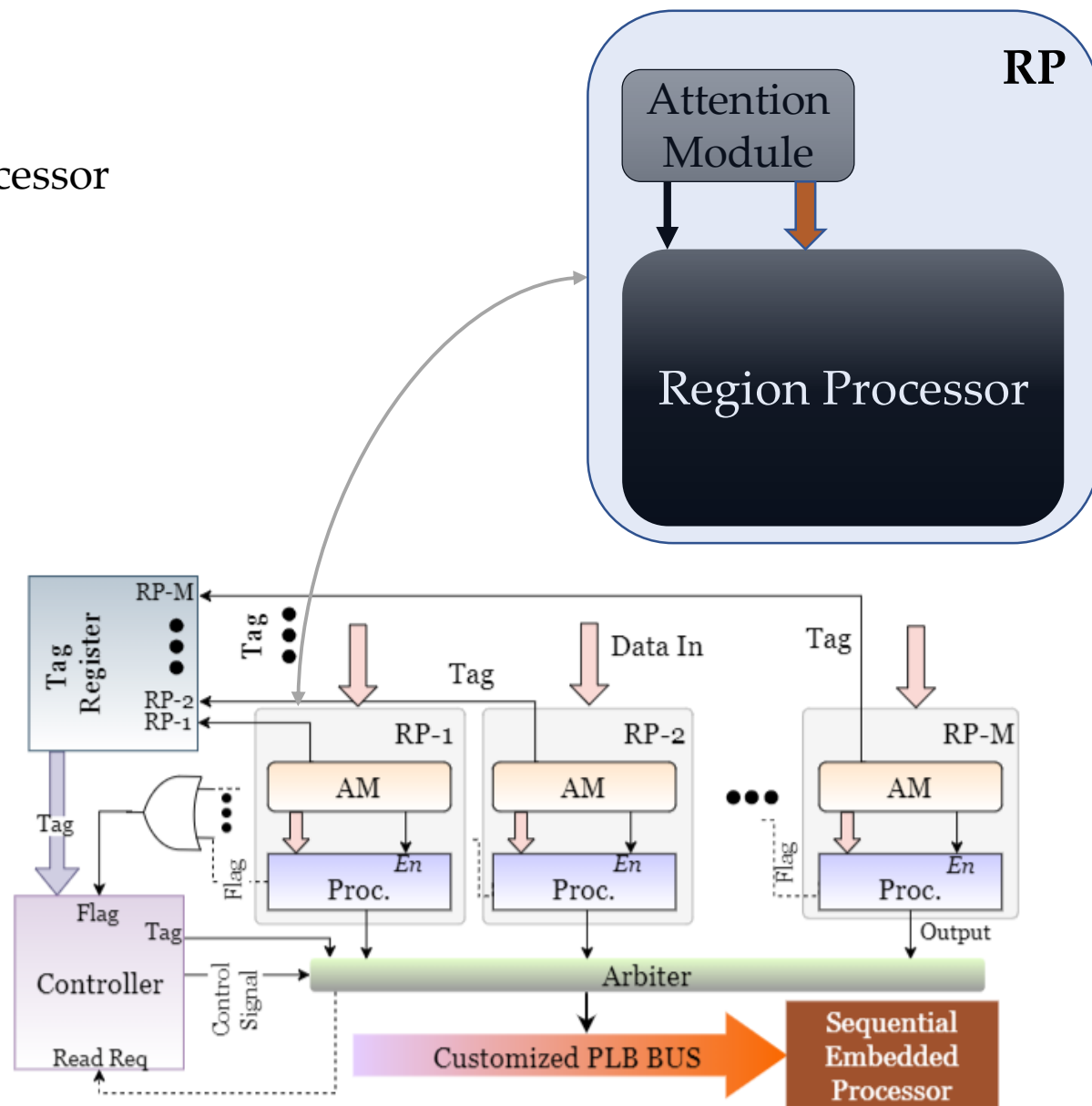
Region-Level Processing Layer

Attention Module (AM):

✓ Computes spatial saliency and activates the processor

Processor

- ✓ RP works on a larger region and provides region-level parallelism.
- ✓ Attention Module removes spatial redundancy.
- ✓ Region processors activates if there is an ROI.
- ✓ Arbiter sends global features to sequential processor from RP-1 to RP-M and only active RPs get the bus access.



Knowledge Inference Processing Layer

- ✓ The external embedded processor receives reduces data from an image after removing temporal and spatial saliency.
- ✓ The received data it pre-processed by low-level and mid-level image processing operations.
- ✓ Hence the remainder operation in the machine vision task is less time consuming and computationally inexpensive.

Result Analysis: Pixel Level

- ❖ **Thirteen** low-level image processing applications are runtime reconfigurable.
- ❖ Predictive-Coding in Time removes temporal redundancy and gives **84.01% power savings**.
- ❖ The area and power overhead are in an acceptable range.
- ❖ The design achieves huge acceleration at the pixel level.

Table 2: Execution Time comparison for executing low-level Applications

Image Size	Execution Time (ET)				
	CPU	Embedded Processor	GPU	FPGA	Our Implementation
160x70	2 ms	28.74 ms	5.47 us	2 us	12.5 ns
320x140	5 ms	115.59 ms	5.50 us	4 us	12.5 ns
640x480	21 ms	791.07 ms	5.53 us	8 us	12.5 ns

Table 1: Accurate prediction saves power from pixel level processors

	Prediction by the AcM		Without AcM
	Inaccurate	Accurate	
Power Consumption (mW)	4.884	0.781	4.1892
Dynamic Power Savings (%)	~	84.01	14.22
Execution Time (ns)	12.5	1.25	11.25
Speed up	~	10×	1.11×

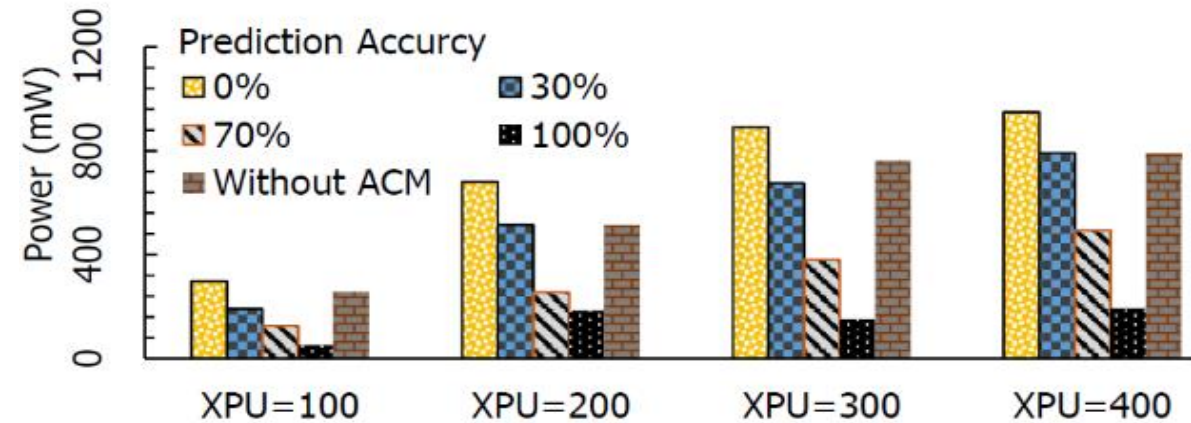


Figure: Possible power savings from different group of processing units (XPU)

Result Analysis: Region Level

RP performs the following operation on an 8x8 (*scalable*) region

- Region growing for segmentation
- Corner detection
- Scale Invariant Feature Detection (SIFT) for object recognition

Key Features

Runtime Reconfigurable for mid-level image processing applications
 Saves **96.91%** power by dynamically postponing execution
 Reduces data volume for sequential processor

Table 3: Parameter Extraction of the modules In ASIC

Module	Area (um ²)	Power (mW)	Leakage (uW)	Delay (ns)
Attention Module	142.84	0.127	2.81	80
Proc.(Corner)	13830.67	11.46	333	93.75
Proc.(SIFT)	13804.86	11.43	331	146.25
Proc.(Seg)	886.3	3.938	43	160

Table 4: Parameter Extraction of the modules In FPGA

Module	Flip-Flop	LUT	No of clo-ck cycles	Power (mW)
Attention Module	422	168	64	15
Proc.(Corner)	768	1349	75	47
Proc.(SIFT)	1180	2323	117	128
Proc.(Segmentation)	661	1060	128	28

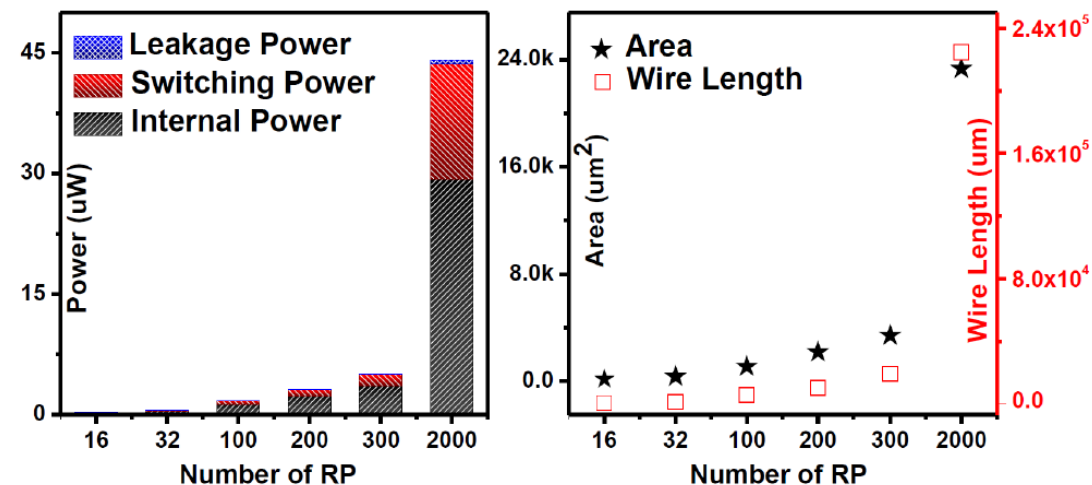


Figure: Increase in resource utilization for different number of PRs in the customized processor local BUS

Summary

- ❑ We prototype inter-pixel processor for an image sensor with hierarchical processing.
- ❑ Pixel-level processors execute low-level image processing applications after removing temporal redundancy.
- ❑ Region-level processors execute mid-level image processing applications after removing spatial redundancy.
- ❑ The gradual degradation of data volume accelerates the remainder task in a machine vision task on the sequential processor.
- ❑ Biological vision system is emulated on those processors to reduce power and speedup the system.
- ❑ The processors are implemented on hardware (ASIC and FPGA) and they are runtime reconfigurable to ensure a flexible design after fabrication.

Thank You
for
Your Concentration