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Enabling Ultra-low Power Machine Learning at the Edge

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### **Event Sensors for Embedded Edge Al** Vision Applications

Christoph Posch, Co-founder and CTO, PROPHESEE





- Introduction event-based vision
- Event sensors for edge-AI applications requirements and design targets
- Design and implementation:
  - Sensor features
  - Processing features
  - Data formats and interfaces
  - Power modes
- Summary / Conclusion

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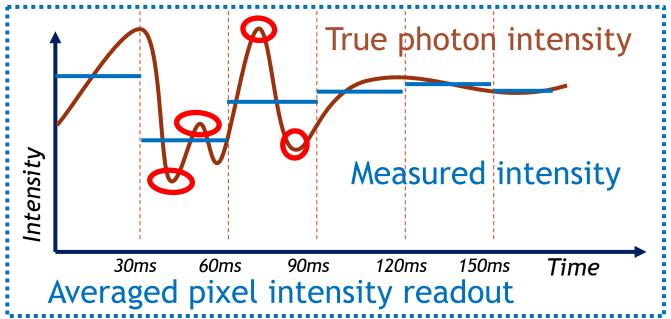
# Image Sensing and Computer Vision

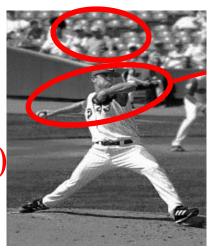
### Image Sensor

- Integrate light intensity in a time interval (exposure time)
- read out average exposure
- at *regular* intervals (frame rate)
- for *all* pixels
- ➔ Image = snapshot in time (static) CV: Motion acquisition based on series of *static* images

### Problems with frames in CV:

- 1. Temporal details lost (under-sampling)
- 2. Displacement, blur (under-sampling)
- 3. Redundant data (over-sampling)





### **Introduction Event-based Vision**

Different dynamics in different parts of the scene

• Each pixel sees a different signal

Not one sampling rate for all pixels (=frame rate)

- but many (= as many as pixels), and
- sampling rates can **vary**, on the fly, and **pixel-individually**

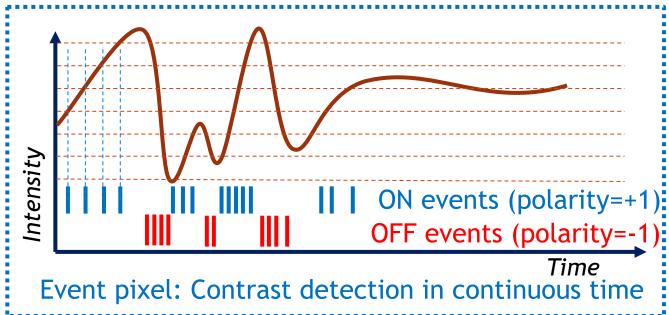
How? Each pixel individually controls its own sampling based on the input signal

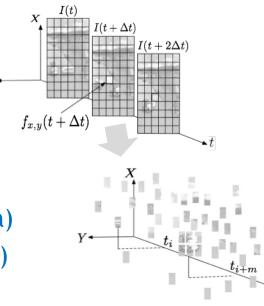
- Change sampling domain → from time to **amplitude**
- Encode information in "events"

From images to stream of pixel-individual data

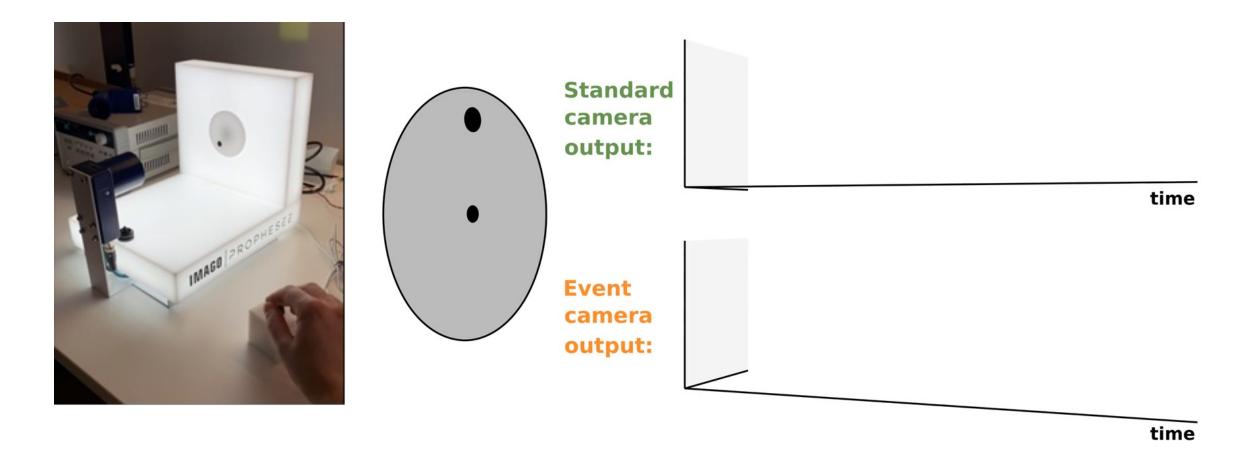
#### **Event sensor characteristics**

- Captures fine temporal detail of motion  $\rightarrow$  fast
- Inherent data compression by eliminating redundant information  $\rightarrow$  sparse (focus on <u>relevant</u> dynamic data)
- Pixel-individual log-domain operation (relative change)
  → high dynamic range





### Introduction > Event Sensor Acquisition Principle



UNIZH, Robotics and Perception Group, https://rpg.ifi.uzh.ch/

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## **Event Sensors for Edge-Al Applications**

### Challenges for event sensors usage and integration

- Diverse fields of application for event sensors: Industrial, surveillance, IoT, AR/VR, mobile, automotive, ...
- Unconventional format of the event data
- Unfamiliar encoding of dynamic visual information as events
- Non-constant data rates
- Non-standard interfaces and data formats, non-standardized interface protocols

Prophesee has developed the first of a new generation of event sensor designed with the explicit goal to improve integrability and usability of event sensing technology in embedded at-the-edge vision system.

# **Design Targets**

### Integrability and usability in embedded edge-AI and IoT vision systems

### • Interfacing and system integration

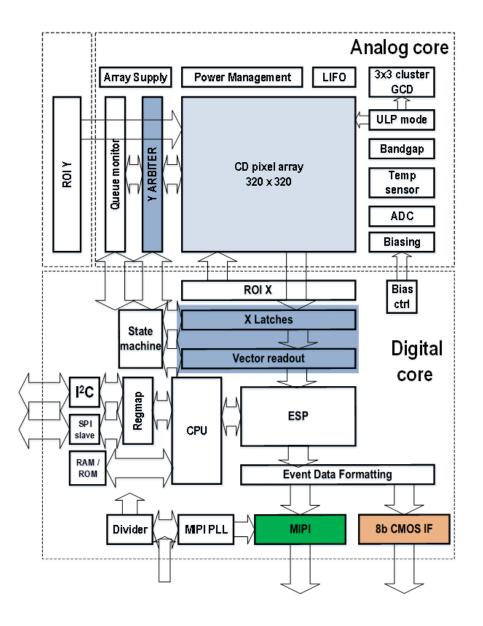
- Flexible/programmable event data pre-processing, filtering and formatting
- Industry standard interface compatibility (MIPI, DCMI)
- Low-latency connectivity for low-power uCs, neuromorphic processor architectures, SNN accelerators, ...

### Power optimization

- Hierarchy of power-modes for adaptative application-specific power optimization
- Ultra-low power/always-on operation with sensor and system wake-up features
- On-chip power management
- Embedded microcontroller core for improved sensor flexibility and useability atthe-edge (data statistics, meta data, microcode execution)

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# **Sensor Chip Overview**



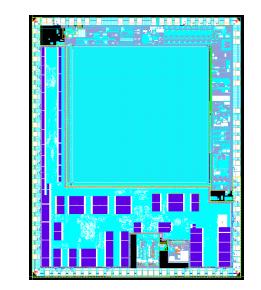
- 320x320 pixels, 1/5" optical, 13mm<sup>2</sup> die
  - Cu-Cu Stacked BSI, (CIS 65nm, CMOS 40nm)
  - Pixel pitch: 6.3um

### Sensor features:

- Global Contrast Detector (GCD)
- Pixel-level Flexible Region-of-Interest (ROI)
- Temperature sensor
- Ambient light sensor
- ULP ultra-low power modes

### • Processing features:

- Digital ESP (Event Signal Processing) pipeline with
- Noise/Flicker Filters, Edge enhancement, ERC, EDF
- Embedded RISC-V CPU
- Data interfaces: MIPI (D-PHY), CMOS Parallel (DCMI, AER)
- Configuration interfaces: SPI/I<sup>2</sup>C

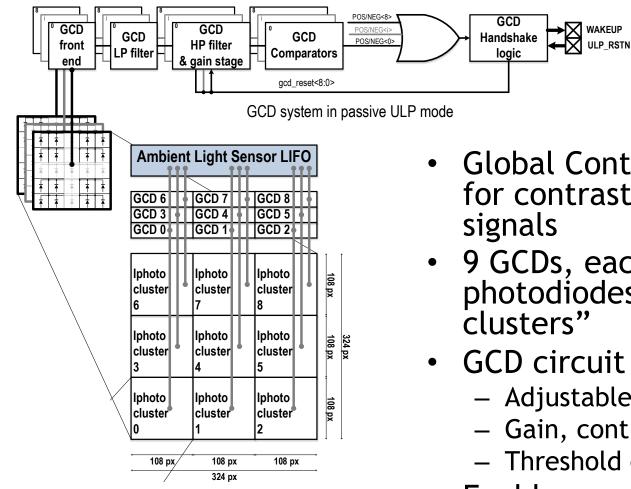


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### – Sensor features

- Processing features
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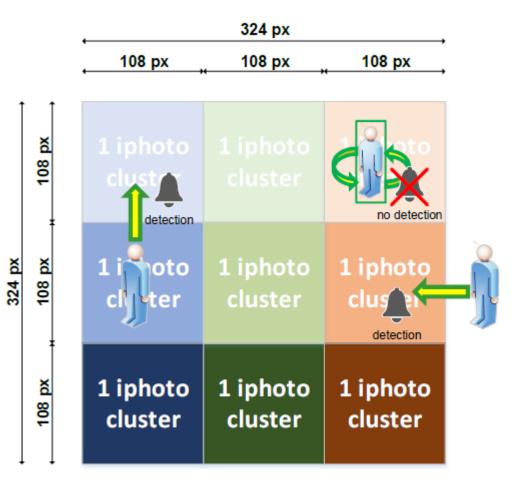
### Sensor Features > Global Contrast Detector



- Global Contrast Detector (GCD)  $\rightarrow$  circuit block for contrast detection from multiple photodiode
- 9 GCDs, each connected to 108x108 photodiodes, forming a 3x3 matrix of "pixelclusters"
- GCD circuit features:
  - Adjustable low/high-pass filter
  - Gain, contrast sensitivity controls
  - Threshold comparators
- Enables smart motion detection at ultra-low power consumption

### Sensor Features > Global Contrast Detector

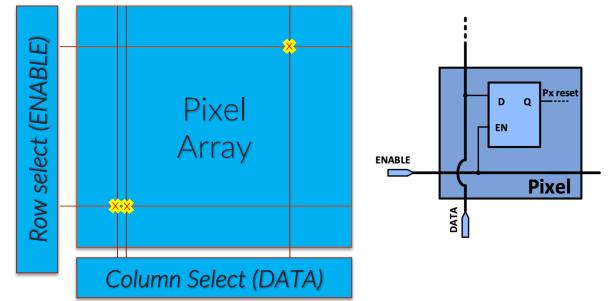
- GCD detect moving object entering/exiting pixel-cluster's FoV
- Enables ultra-low power / always-on smart scene activity detection
- Create internal and external (chip pin) wake-up signals in Passive Ultra Low-Power mode.

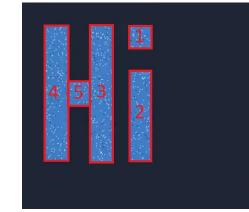


#### Passive ULP mode

### Sensor Features > Pixel-level Flexible ROI

- All pixels embed in-pixel latch for activation/deactivation
- Each latch is individually controllable
- Real-time operation
- Usages:
  - Hot pixel removal
  - Fully customizable ROI shapes
  - Dynamic saliency
  - Limit power consumption outside ROI
- Windows mode ROI controller:
  - Ease user programming of multiple rectangular pixel array regions
  - Speed-up programming for dynamic usage
  - Manages automatically up to 18 ROI windows

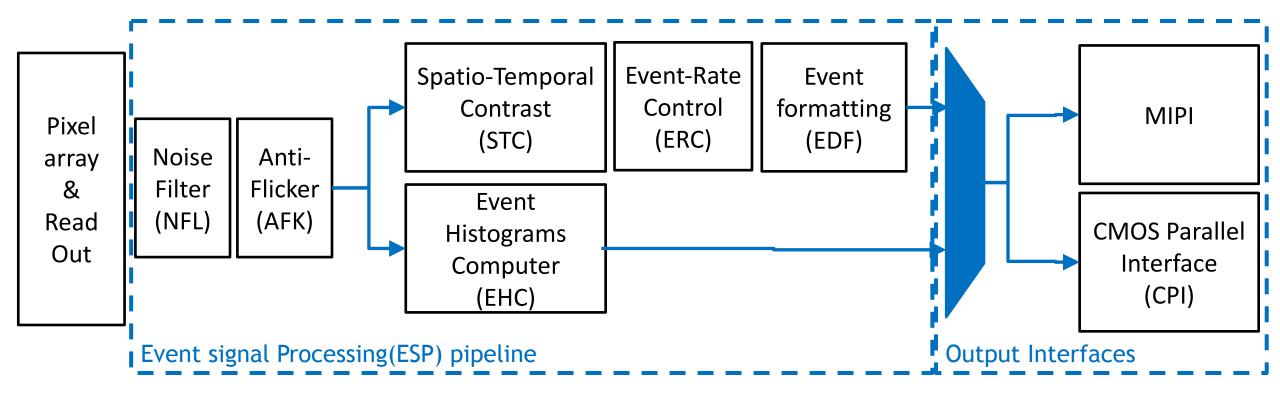




Pixel array with 5 ROI Windows programmed

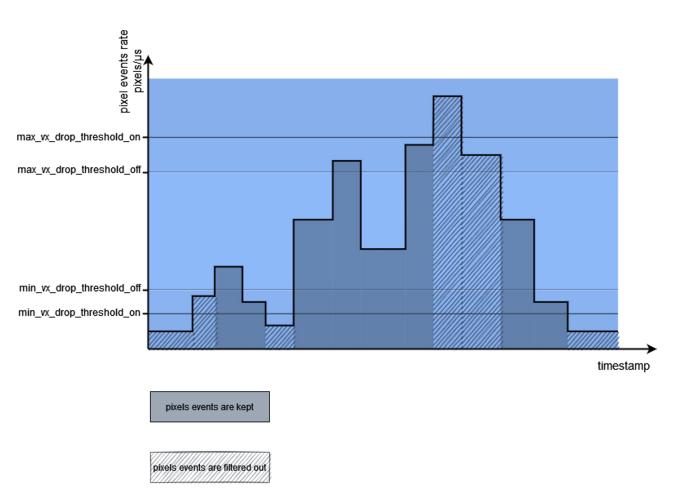
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### Processing features > ESP-Event Signal Processing



### Processing features > Noise Filter (NFL)

- NFL removes events according to instantaneous event rates during a sliding time window.
- Programmable thresholds are used to determined if the NFL is dropping or passing incoming events
- Thresholds are programmable separately for the two event polarities



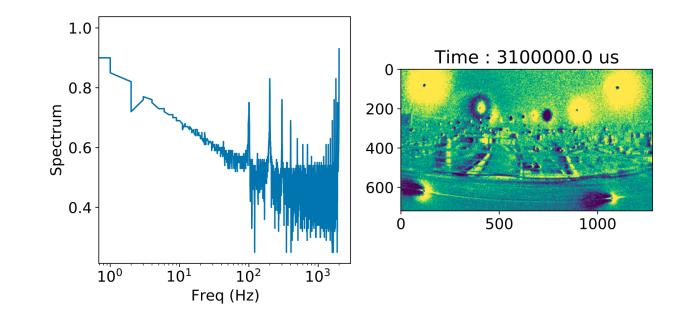
## Processing features > Anti-Flicker Filter (AFK)

Natural scenes often contain light sources flickering at constant frequencies: Fluorescent lights, LEDs, screens, ...

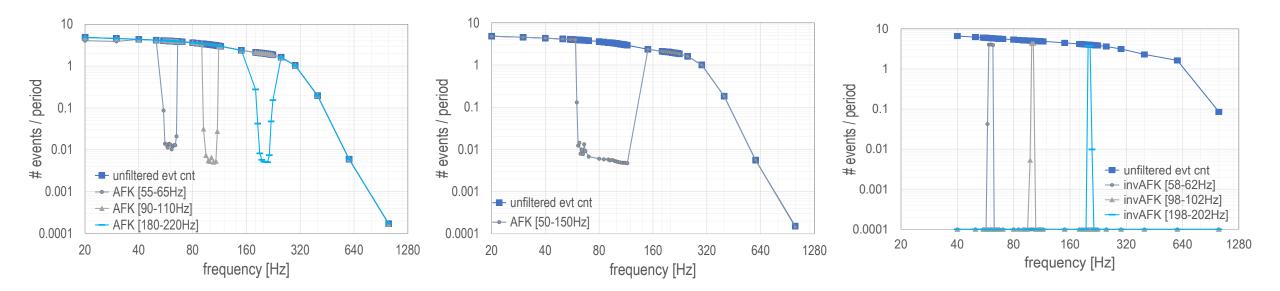


### **AFK requirements:**

- Digital band stop (notch) filter
- Programmable center/width
- Frequency Range 50-500Hz
- Invertible (band pass)



### **Processing features > AFK Measurements**



#### Band stop (notch)

- Center frequencies: 50Hz, 100Hz, 200Hz
- Bandwidth ±10% of center
- Typical attenuation ~50dB

#### Band stop

- Center frequency: 100Hz
- Bandwidth 100Hz
- Attenuation >50dB

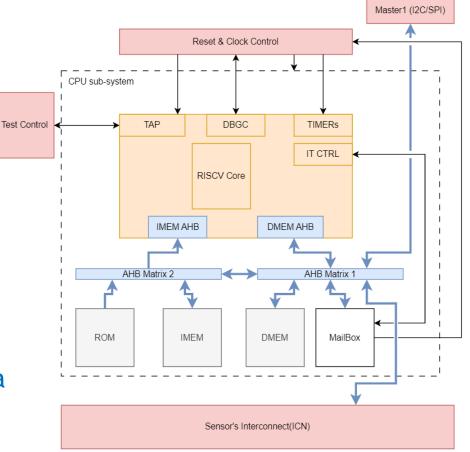
#### Inverted (band pass)

- Center frequencies: 50Hz, 100Hz, 200Hz
- Bandwidths ±2Hz of the center

# Processing features > Embedded CPU (RISC-V)

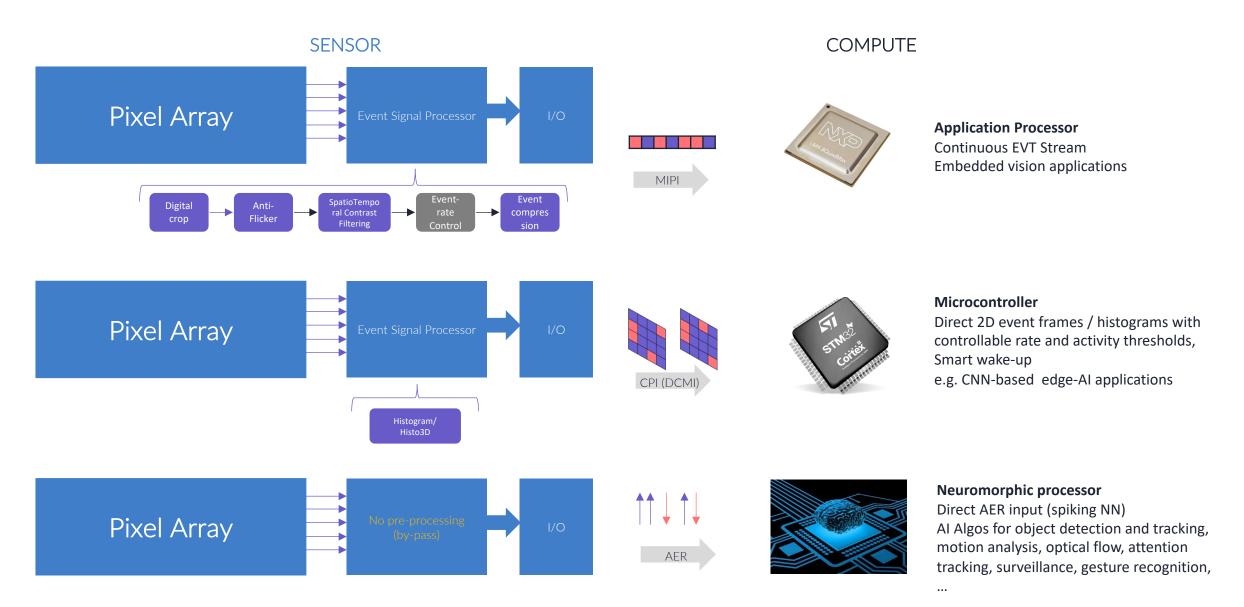
### **RISC-V RV32IMC ISA**

- Interrupt controller, 16 IRQ lines
- Peripherals:
  - 3 x 64-bits Timers; Real Time Clock (1MHz)
  - Mailbox system for host communication
  - JTAG interface, HW breakpoint
- Usage:
  - Boot Initialization (Time to First Data)
  - Power manager (entering/exiting power modes)
  - MIPI meta data insertion (e.g. frame number, data statistics, etc...)
  - (No algo-based event data processing)



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### Event Sensor → Compute Platform



#### PROPHESEE

### **Data Formats**

#### **Event streaming**

- EVT2.0: legacy, low event rate, 32-bit per pixel, uncompressed, timestamp
- EVT3.0: compressed vector 16-bit encoding, high event rate, removes redundant event data, best encoding on average
- EVT2.1: vectorized, high event rate, 64-bit based
- AER: Address Event Representation format, legacy format used by the first event-based sensors. Low latency, no timestamps

### **Event accumulation**

- Histograms: Accumulating events into a 2D grid structure
  - fixed time window
  - fixed number of events
- Accumulation methods:
  - Histo3D: 2 separate containers per pixel (positive, negative events)
  - **Diff3D:** 1 container per pixel (add positive, substract negative events)

Data formats	CPI @50MHz		MIPI @1.5Ghz	
	4 bits Mevt/s	8bits Mevt/s	Mevt/s	
EVT2.1	100	200	750	
EVT2.0	6.25	12.5	46.87	
EVT3.0	133.33	266.67	1000	
AER	10	16.67	N/A	

### **Output Data Interfaces**

#### AER

- X, Y, polarity only (no TS)
- Low latency (no buffering)
- Variable data rate

#### TS Event Streaming

- X, Y, polarity, TS
- Variable data rate
- Lossless Compression using EVT format

#### **3D Histograms**

- X, Y, polarity accumulated
- time period or number of events
- Fixed size, NCHW data org (for NPU DMA)

### **CPI PARALLEL INTERFACE:**

- > 4/8-bit CMOS parallel
- > Up to 400Mbps @50Mhz (267Mevt/s)
- > **Low-power** (PLL not required, clock can be adjusted for different bandwidth reqs)
- > Low end-to-end latency (very little buffering)
- > Different transmission modes:

Protocol	Pin config	Output Data format		
		AER	EVTx	Histogram
PSEE-IF	4/8bits	Х	Х	
DCMI-JPEG	8bits	Х	Х	
DCMI-Mono	8bits		Х	Х
AER-IF	4/8bits	Х		

#### **MIPI INTERFACE:**

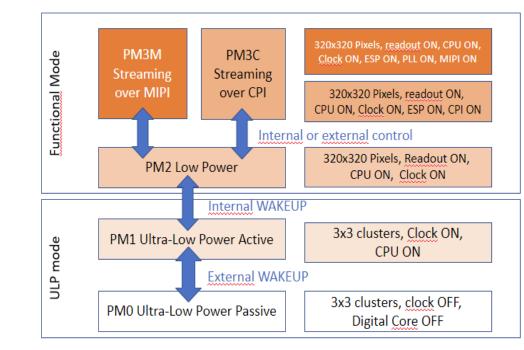
- 1 lane, up to 1.5Gbps @1.5Ghz (1Gevt/s)
- Compliant with MIPI D-PHY V1.2
- Supported formats:
  - Compressed/uncompressed vector formats (EVTX.X)
  - Event histograms
  - Frame size, frame rate configurable
- Optionally fixed frame size/fixed frame rate with data padding to comply with standard configuration for FB cameras.

#### Low-power / Low-latency

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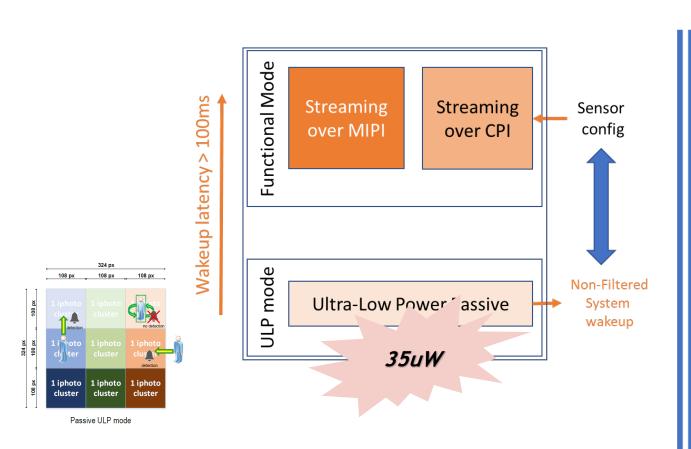
### **Power Modes - Overview**

- **PMO**: Digital Core is OFF, activity detection based on Analog GCD
- PM1: Digital CPU-only is ON, activity detection with advanced CPU processing on GCD data
- **PM2**: Digital Core is partially ON, activity detection with advanced CPU processing on pixel data, no streaming
- **PM3C**: All ON (except PLL), full streaming on CPI interface
- **PM3M:** All ON, full streaming on MIPI interface



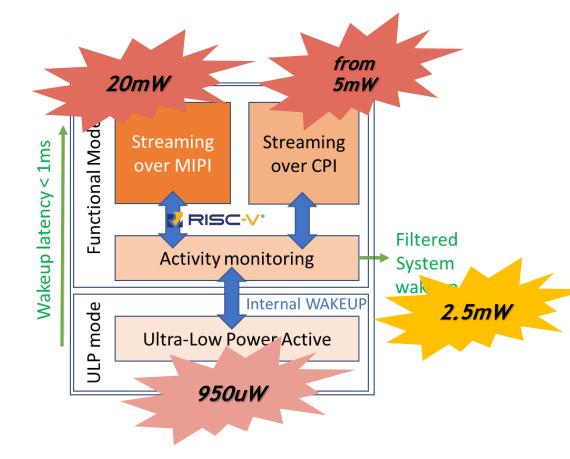
Mode	Ultra Low Power passive PM0	Ultra Low Power active PM1	Low Power PM2	CPI streaming PM3C	MIPI streaming PM3M
Sub-system			100kEPS	100kEPS CPI @10MHz	100kEPS 800MHz
Pixel array	3x3 GCD resolution	3x3 GCD resolution	Full resolution	Full resolution	Full resolution
Digital Registers+CPU	Power off	powered, clocked	powered, clocked	powered, clocked	powered, clocked
Digital readout	Power off	powered, clock gated	powered, clocked	powered, clocked	powered, clocked
Digital ESP + Output IF	Power off	powered, clock gated	powered, clock gated	powered, clocked	powered, clocked
	Analog: 35µW Digital: 0	Analog: 35μW Digital: 915μW	Analog: 1.2mW Digital: 1.3mW	Analog: 1.2mW Digital: 2.3mW	Analog: 1.2mW Digital: 5mW MIPI+PLL: 10mW
	Total: 35µW	Total: 950µW	Total: 2.5mW	Total: 3.5mW	Total: 16.2mW

### **Power Modes and Wake-Up Features**



#### Wake-up on events from 3x3 clusters

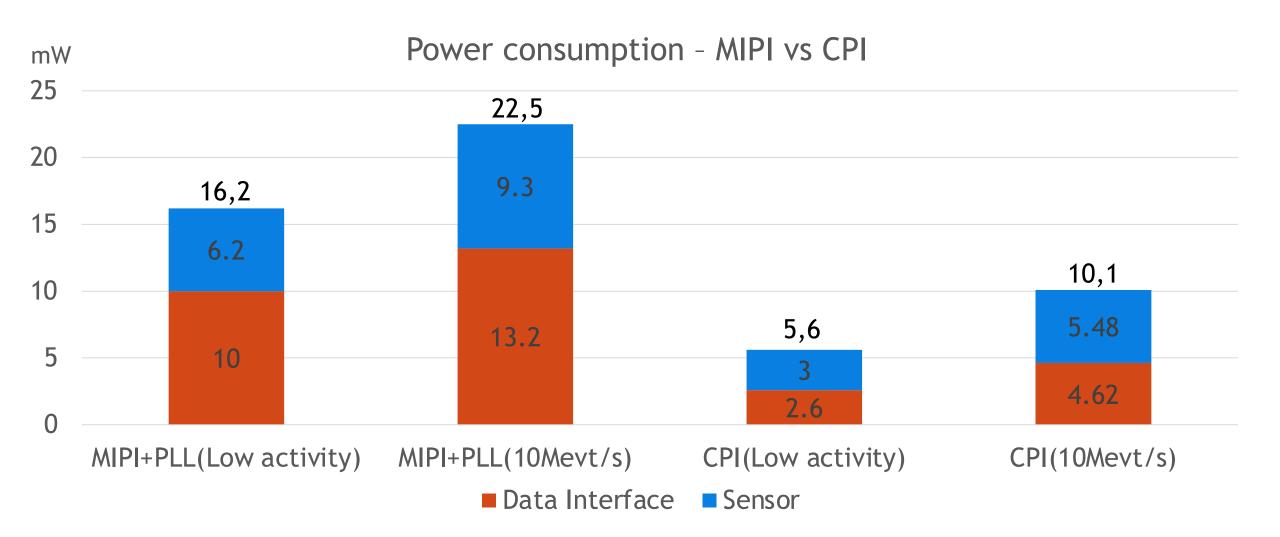
- Internal: Power-mode cycling
- External: system wake-up signal on pin



# PM2: Activity monitoring in 4x4 programmable zones

- Tuneable thresholds
- CPU statistical processing for wake-up filtering (false positive)

### Power Modes > Streaming Modes: MIPI vs CPI



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# Summary / Conclusion

- Specifically designed for embedded vision at-the-edge
- Cu-Cu Stacked BSI CIS on CMOS (65nm/40nm)
- Features multiple data pre-processing, filtering and formatting functions to adapt to variety of use cases and processing platforms
- Two data interface covering low latency applications (CPI) and high bandwidth applications (MIPI)
- Multiple transmission options in both interfaces ease integration and compliance to industry standards
- Adaptative hierarchical power modes facilitate operability in power-sensitive edge vision applications
- On-chip power management and an embedded uC core further improve sensor flexibility and useability at-the-edge

### Thank you for your attention



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