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Low-Power Event-driven Image Sensor Architectures

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Internet of Things Challenges



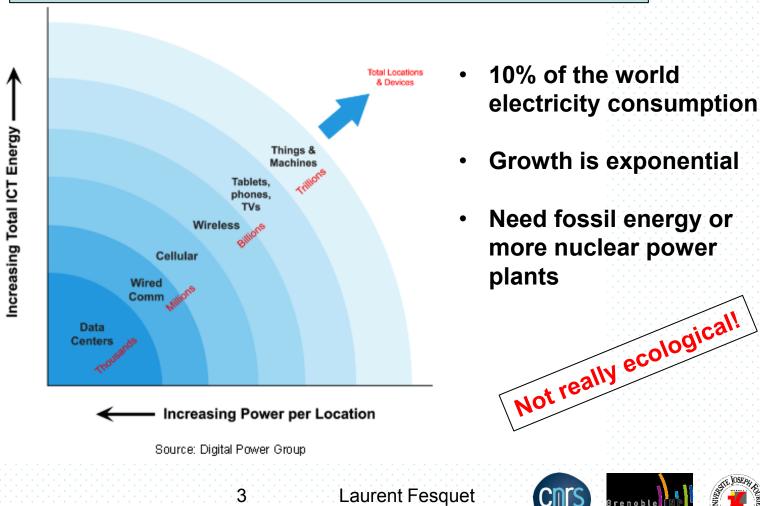






Upcoming Challenges

Where electricity is consumed in the digital universe?





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Outline

Another way for sampling

- Sampling without Shannon?
- Nonuniform Sampling

Sampling an image

- State of art: Conventional Image Sensors
- State of art: Event-based Image Sensors
- An architecture for low-power Image Sensors

Conclusions

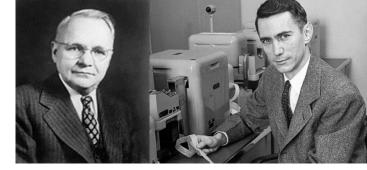








Important questions



Harry Nyquist and Claude Shannon

Claude Shannon and Harry Nyquist are they responsible of the digital data deluge?

• We will not answer to this question but the big data is today a reality!

Can we find a better sampling scheme to stem this digital data deluge and stop the energy waste?

• We hope so !!!



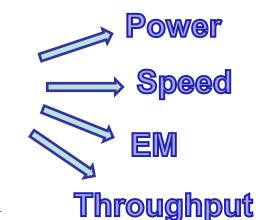






Sampling is the success key

- Sampling based on the Shannon-Nyquist theorem
 - Efficient and general theory... whatever the signals!
- Smart sampling techniques
 - More efficient but less general for specific signals!



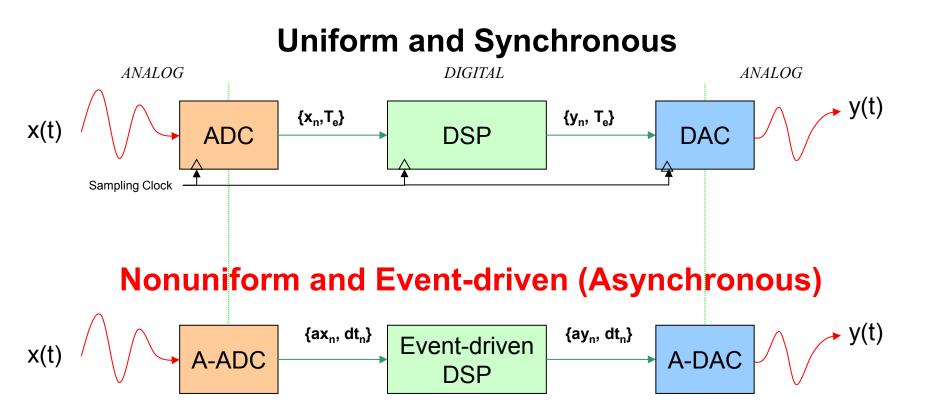
- Need a more general mathematical framework
- F. Beutler, "Sampling Theorems and Bases in a Hilbert Space", Information and Control, vol.4, 97-117,1961
- Sampling should be specific to signals and applications







What can we do?



Direct processing of the nonuniform samples









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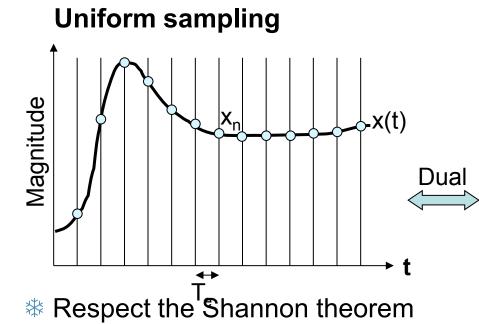








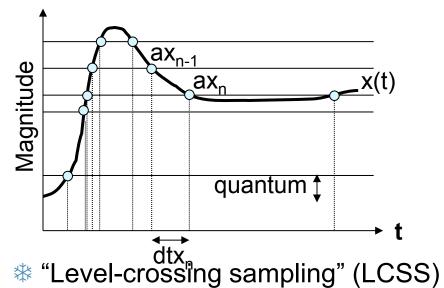
Differently sampling



- Instants exactly known
- * Information: T_{sample} , { i_k }
- In an ADC: Amplitude quantization In an A-ADC: Time quantization
- Many useless samples



Non uniform sampling



- Amplitudes exactly known
- Information: quanta, { dti_k }
- Only useful samples







SNR with non-uniform sampling

• SNR for a sinusoid:

 $SNR_{dB} = 1,76 + 6,02.N$ \implies Number of bits

- Theoretically, noise is only due to amplitude quantization
- With non-uniform sampling, the time is quantized

$$SNR_{dB} = -11, 2 - 20\log(fT_c)$$
 Timer period

 Noise only depends on the timer resolution whatever the threshold distribution



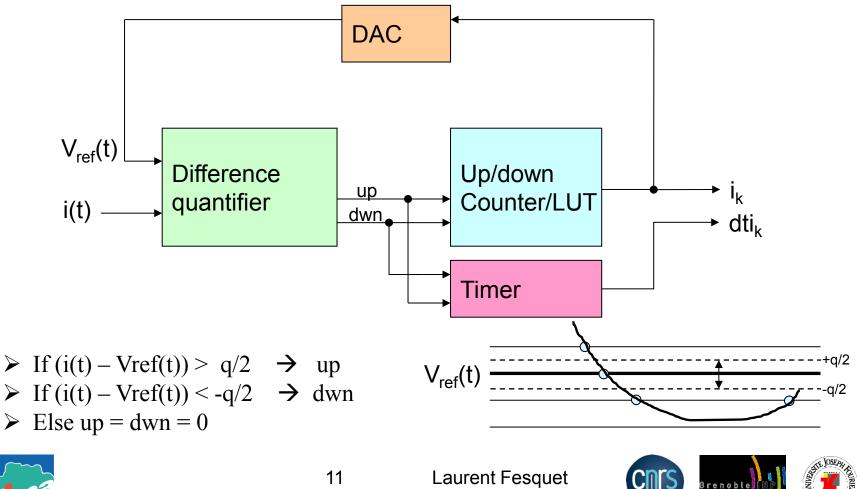






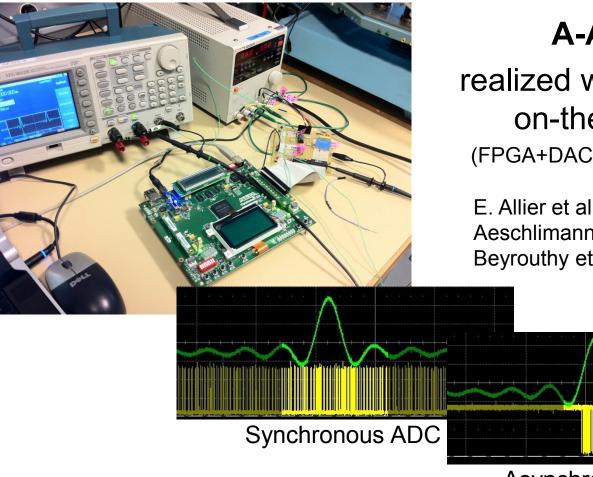
A-ADC or ADC for LCSS

A-ADC for Non Uniform Sampling





A-ADC



A-ADC

realized with circuits on-the-shelf. (FPGA+DAC+Comparators)

E. Allier et al. 05 Aeschlimann et al., 06 Beyrouthy et al., 11

Asynchronous ADC



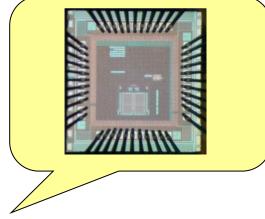




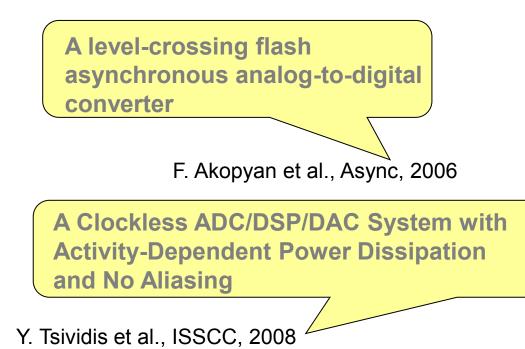


A-ADC testchips

Microphotography of the A-ADC In CMOS 130 nm technology from STMicroelectronics



E. Allier et al., Async, 2003



Lowering in one step the **storage**, the **processing**, the **communications** and the **power consumption!**





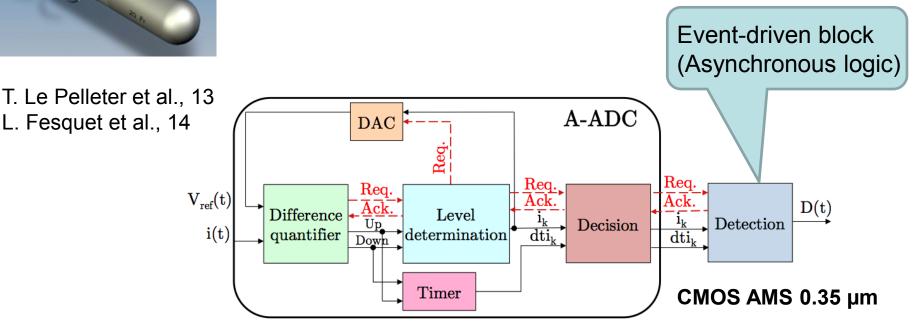


A successful experiment





Activity patient measurements



Experiments based on real physiological signals (recorded on the patients)







What we learned with this experiment

With the medical implant

- No pre-processing
- Less than 1% of data compared to the uniform sampling
- 3 orders of magnitude reduction on power

In general

- **Be more specific** to signals and applications
- Non-general approach, but reproducible
- Non-uniform sampling well-suited to sporadic signals







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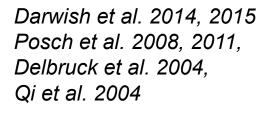
Low-power Image Sensors? Objective of this presentation

- No image sensor really dedicated to low-power
- High resolution sensors requires a high speed ADC
 - High power consumption
- How to reduce the power in Image Sensors

Apply nonuniform sampling in 2D

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- Use advanced technology nodes (expensive)
- Reduce the data flow / activity
- Rethink analog-to-digital conversion











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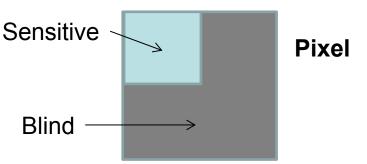






Image sensor principles

- Based on photodiodes
- Pixel fill factor = optical quality
- All pixels are read in sequence
- Larger the sensor, higher the throughput (fixed frame rate)
- Higher the throughput, higher the ADC consumption
- The ADC is the first contributor of power consumption



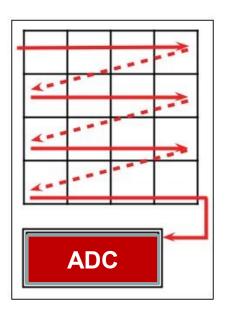
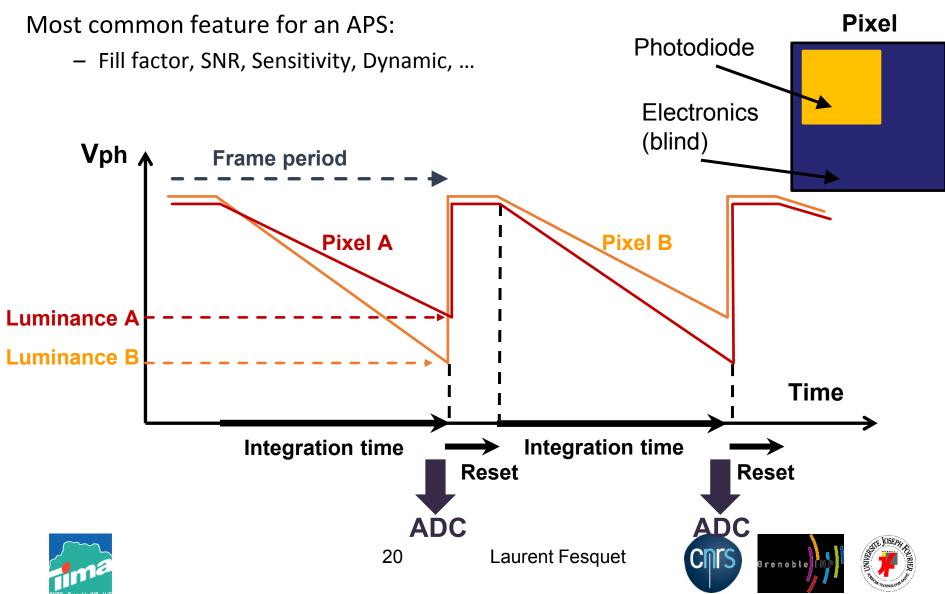








Image Sensor Principles



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Event-based Image Sensors

Pixel reading on event

- Initiated by the pixel itself
- Frameless sensor
- No frame rate
- Address Event Representation (AER)

Event-based pixel

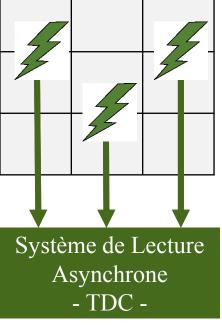
- Asynchronous communication
- Time-to-First Spike encoding (TFS)
- Time-to-Digital Conversion (no more ADC)
- Sensor activity depends on the captured scene
- Reduced dataflow







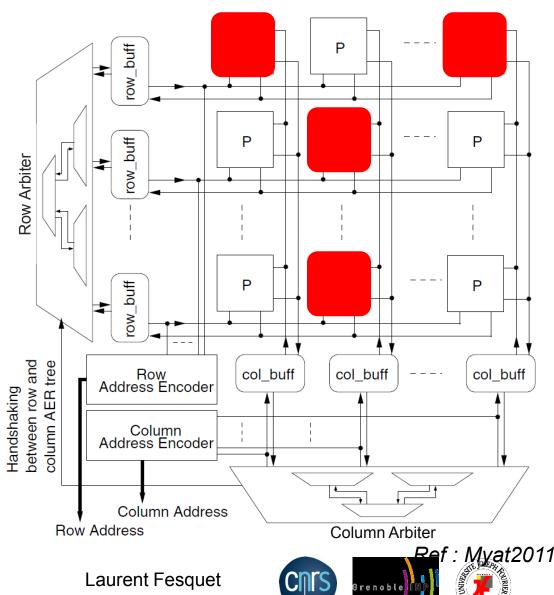




Event-based Image Sensors

- Need to manage multiple pixel requests
- Event-based Image Sensors need arbitration
- Complex Circuitry
 - Exponentialy growth with the image sensor resolution
- Inaccurate time stamping
- Fixed priority arbiters are unfair

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Time-to-First-Spike a bio-inspired approach

- Time represents the information
 - Spike time stamping
 - Time quantization

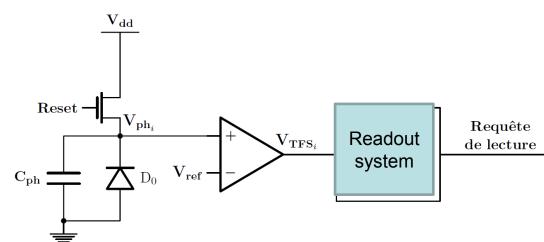
Advantages

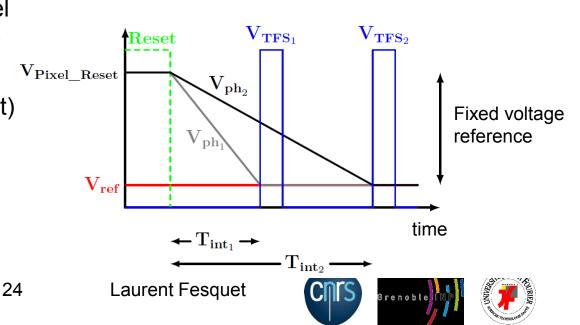
- Read on events
- One integration time per pixel
- Dynmaic range controlled by the TDC
- Robust (digital readout circuit)

Drawbacks

- Large pixel area
- Variability of the analog part
 (comparator)

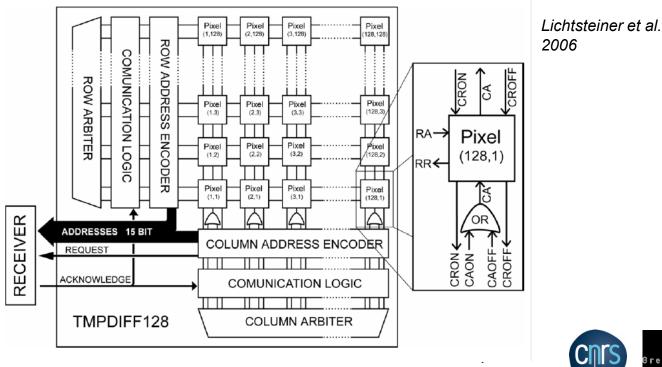






Delbrück/Lichtsteiner : Dynamic Vision Sensor (DVS)

- Neuromorphic approach
- Detect changes in pixel luminance (send a request)
- High dynamic range sensor (logarithmic pixel)

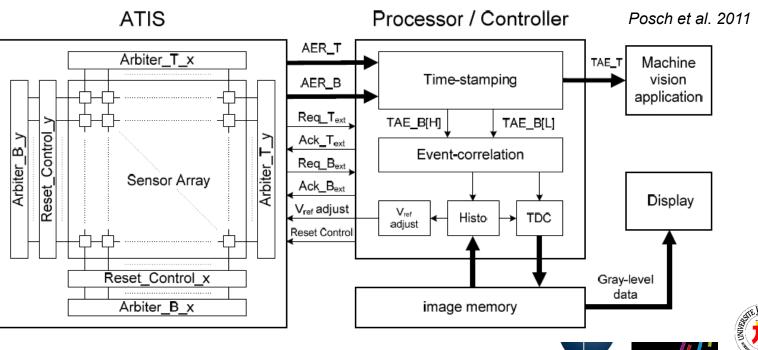






Posch :

- Double arbiter
- Temporal redundancy reduction
- High dynamic range(~ 140 dB)
- large pixel area (89 transistors, 2 capacitances, 2 photodiodes)
- Fill factor < 14 %



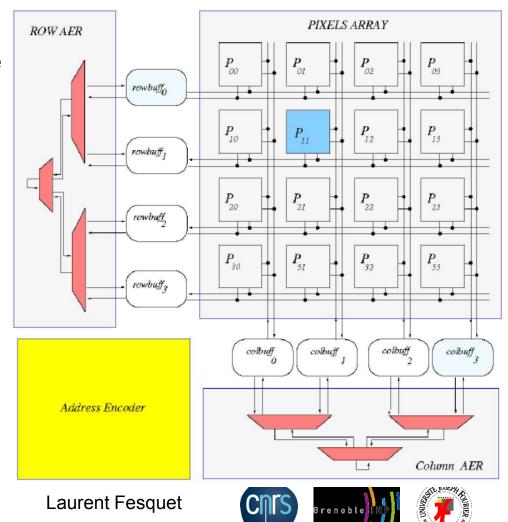


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Shoushun, Bermak 2006

Bermak / Shoushun :

- **Pipelined arbiter** with alternate priority
- Local pixel reset
- Pixel is idle after reading
- Low power design
- Simple pixel design





Parameters	Conventionnal Image Sensors	Event-based Image Sensors	
Sampling Scheme	Uniform	Nonuniform	
Integration Time	One for the whole sensor	One per pixel	
Readout System	Sequential	Event-based	
Protocol	Sequential and Exhaustive	Sequential and Event-based (AER)	
Data Flow	Fixed and Redundant	Nonuniform and Reduced	
Pixel Complexity	Simple	Complex	
Power Consumption	Sub-optimal (redundancy)	Reduced	
Frame Rate	Fixed	Event-dependent	







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Changing the paradigm in a realistic manner

- Keep the fill factor reasonable (important for industrial products)
- Reduce the throughput without changing the frame rate
- Remove the ADC to limit power consumption
- Replace it by a digital circuit (more easy to implement)







Changing the paradigm in a realistic manner



Prefer Time-to-Digital conversion

Use Event-Driven logic (asynchronous)

- Global Clock is *replaced* by local channels (handshaking)
- Power is only consumed when data are really processed

Darwish et al., 14

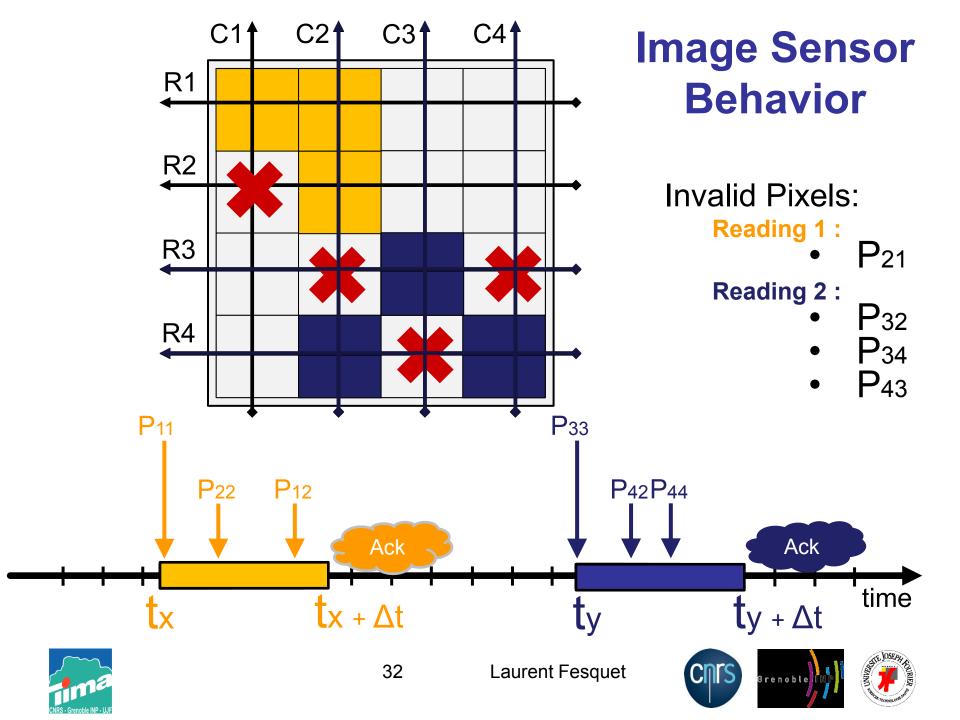
Spatial Redundancies







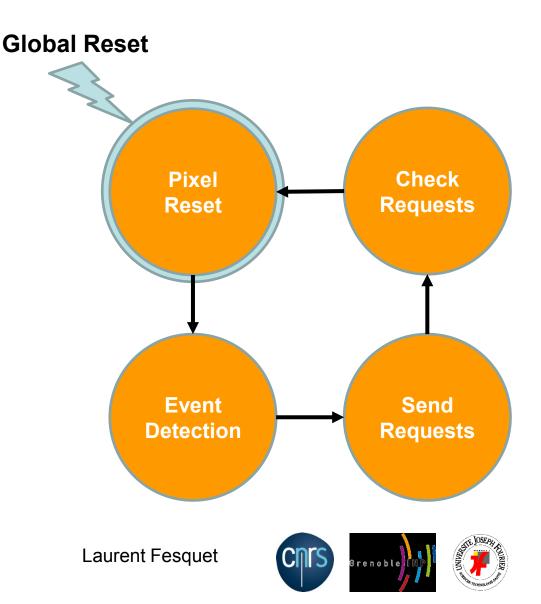
Posch et al., 08 Temporal Redundancies



High-level Pixel Model

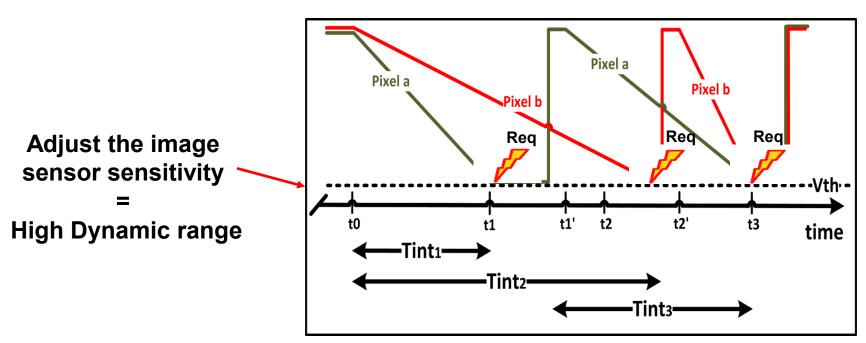
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- 4-phase pixel
 - Event detection
 - Send requests
 - Check requests
 - Reset pixel





Pixel behavior



- Based on event-detection
- 1-level crossing sampling scheme
- Unique integration time per pixel
- Time to first spike encoding

A. Darwish et al., EBCCSP, 2015 A. Darwish et al., NewCAS, 2014





Specific behavior of our sensor

Discussion on Δt

- Δt : deadtime after time stamping.
- No other time stamping allowed during Δt even if events occured
 - Control the time stamping resolution
 - Equivalent to ADC quantum ΔV
 - Control the maximal number of readings or greyscale levels
- Trade-off between image quality and readout rate









Performance Evaluation

- Asynchronous Image Sensor Readout Rate (AISSR)
 - Compute the read pixel number by our architecture (M columns x N rows)
 - Reading operations bounded by the number of luminance values
 - d is the image dynamic range
- PSNR (Peak Signal to Noise Ratio)
 - Evaluate the distorsion induced by the compression
 - For grayscale image, 30dB et 50dB are prety goo values

$$PSNR_{dB} = 10 \cdot \log\left(\frac{d^2}{MSE}\right)$$

 $AISRR(\%) = 100 \cdot \left(\frac{d}{M \cdot N}\right)$

- MSSIM (Mean Structural SIMilarity)
 - Take into account the human perception (luminance, contrast, image structure)
 - Best is 1, worst is 0





 α



Reference Image Readout rate = 100% MSSIM = 1

Δ = 8



Readout rate = 0.018 % PSNR = 36.29dB, MSSIM = 0.765





Readout rate = 0.13 % PSNR = 40.03dB MSSIM = 0.999

Δ = 16



Readout rate = 0.008 % PSNR = 31.46dB MSSIM = 0.635

$\Delta = 4$



Readout rate = 0.042 % PSNR = 38.1dB MSSIM = 0.919

Δ = 32



Readout rate = 0.004 % PSNR = 22.51dB MSSIM = 0.483



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Reference Image Readout rate = 100% MSSIM = 1

 $\Delta = 8$



Readout rate = 0.012 % PSNR = 28.36dB MSSIM = 0.813



Readout rate = 0.084 % PSNR = 31.1dB MSSIM = 0.987

 $\Delta = 16$

Readout rate = 0.006 % PSNR = 26.24dB MSSIM = 0.605



Readout rate = 0.029 % PSNR = 30.34dB MSSIM = 0.942



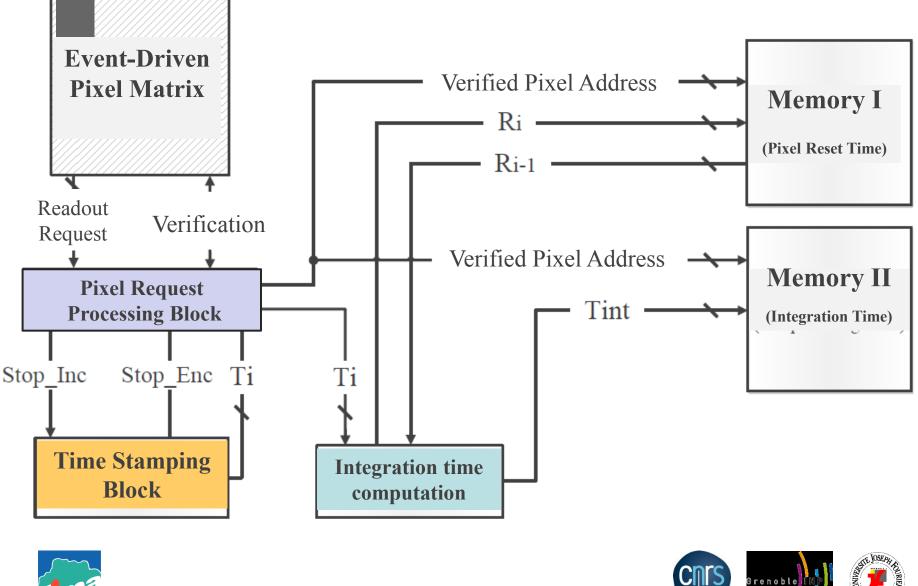
Readout rate = 0.003 % PSNR = 22.13dB, MSSIM = 0.392





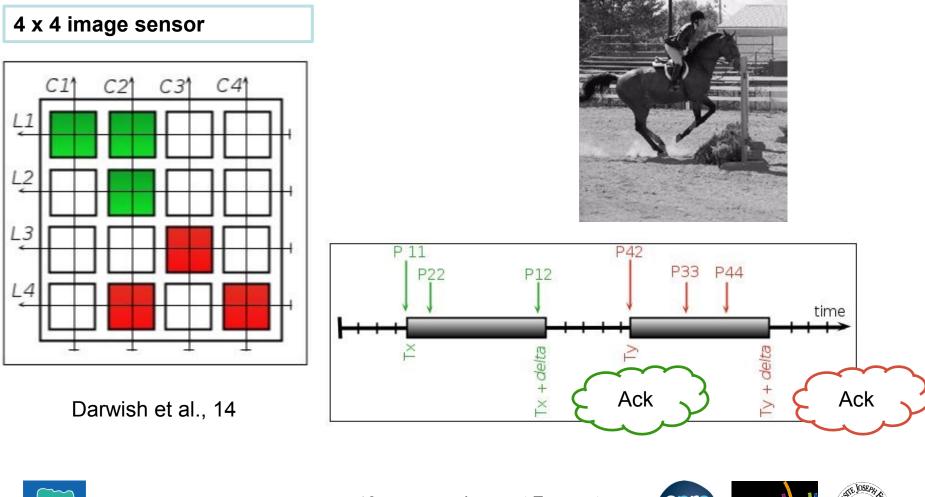


Asynchronous readout architecture





How do we suppress Spatial Redundancy ? (again)





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What we learned

with image sensors

- 1-level crossing sampling in 2D
- Low percentage of readings per column (< 1 %)
- Image data flow reduction
- Event-driven digital circuitry
- Adjustable resolution and dynamic range
- **Don't need an ADC** (power consuming)
- Intrinsic A-to-D conversion







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General Conclusion

- Lesson 1: **Determine** the most efficient sampling!
- Lesson 2: Fit well **Event-Driven Circuits** (asynchronous)
- Lesson 3: Ultra-Low Power
- Less samples means:
 - Less computation, less storage, less communications,
 - less power



A more energy efficient approach of digital processing











Image Sensor Conclusion

Parameters	Delbrück / Lichtsteiner	Posch	Bermak / Shoushun	Our work
Readout Protocol	Sequential	Sequential	Sequential	Parallel
	Non- deterministic	Non- deterministic	Non- deterministic	Deterministic
Arbiters	Yes	Yes	Yes	Νο
Readout rate	Pixel Rate	Pixel Rate	Pixel Rate	Reading Rate
Data Compression	Metadata	Temporal Redundancies	No Compression	Spatial Redundancies
Pixel Size	\odot	$\overline{\mathfrak{S}}$		င္ဗာ to င္လာင္ရွ



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Image Sensor Pespectives

Image sensor fabrication and test



Sampling scheme is **signal-dependent**

- How to directly process the image data flow?
- Smart Image Sensor for Vision and Robotics







Non-uniform sampling is the future of digital universe!



Thanks for your attention





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