

# Event Computer Vision 10 years Assessment: Where We Came From, Where We Are and Where We Are Heading To

R.B. Benosman

University of Pittsburgh, Carnegie Mellon, Sorbonne Universitas

Eye& Ear and McGowan Institute

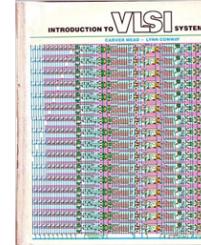
3025 E Carson st,

Pittsburgh, PA 15203



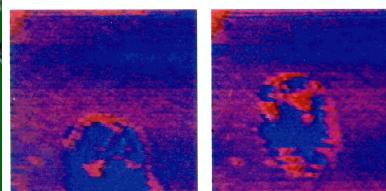
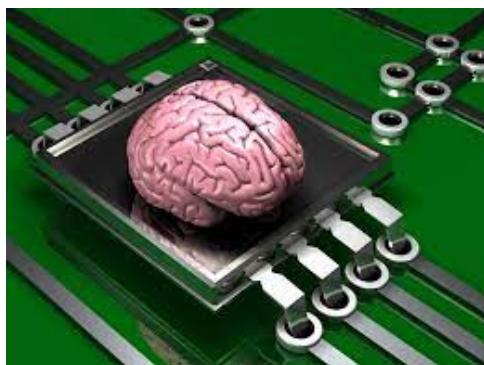
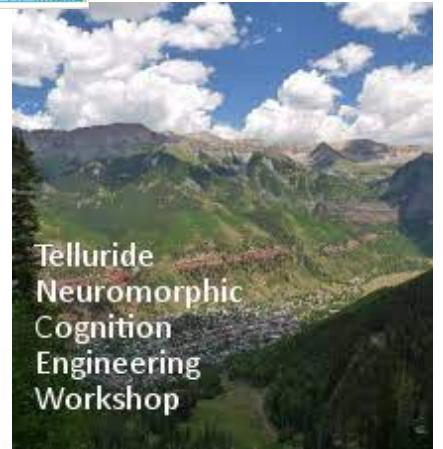
University of  
**Pittsburgh** Carnegie  
Mellon  
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UNIVERSITÉS

# Where we came from...



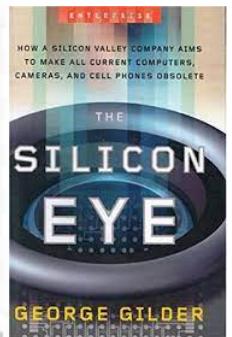
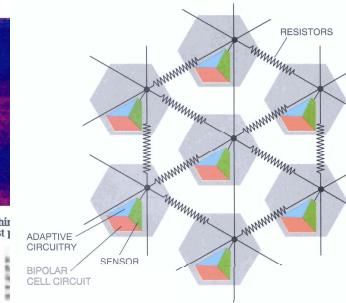
SCIENTIFIC  
AMERICAN

Exploring the genetic heritage of eukaryotes.  
Can anyone explain high-temperature superconductivity?  
The impact of Kuwait's burning oil wells.



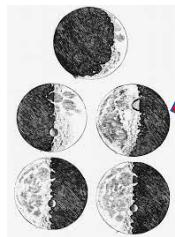
SOCERBALL in motion shows how the delayed response of the horizontal cell network affects the retina's perception. The ball leaves behind spots have just 1

Mahowald/Mead retin

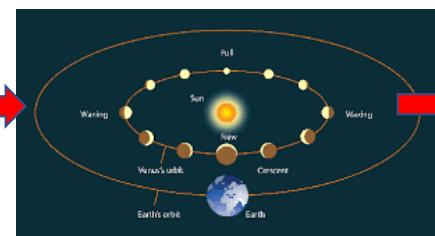


# Where we came from...

## A Chance to Move Perception from Engineering to Basic Science!



observe



understand

### Relativistic orbital motion

- Light and bodies move on geodesics in the spacetime
- The 3D "shadow" of a geodesic in 4D is the orbit
- We need an equation for "straight lines" in spacetime!

$$\frac{d}{d\tau} \left( \frac{\partial \mathcal{L}}{\partial \dot{\sigma}^i} \right) - \frac{\partial \mathcal{L}}{\partial \sigma^i} = 0$$

Euler-Lagrange equation

$$\mathcal{L} = \frac{1}{2} g_{ij} \dot{\sigma}^i \dot{\sigma}^j$$

The metric tensor  $g$  describes spacetime curvature, contains the Schwarzschild metric coefficients

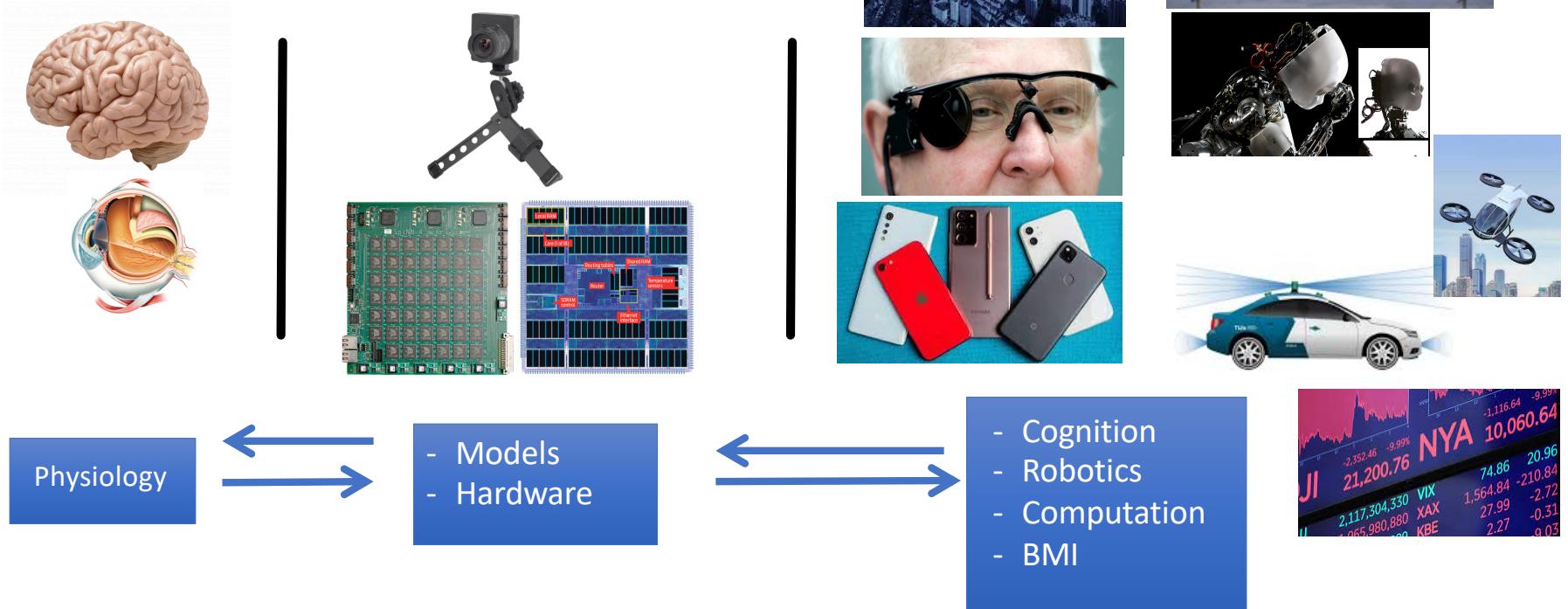
model



application



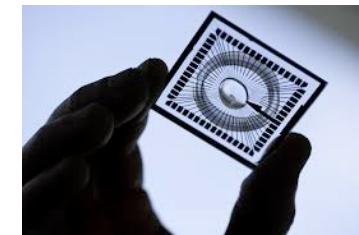
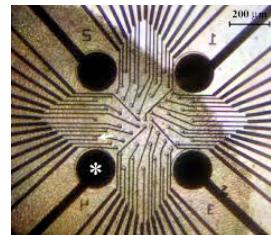
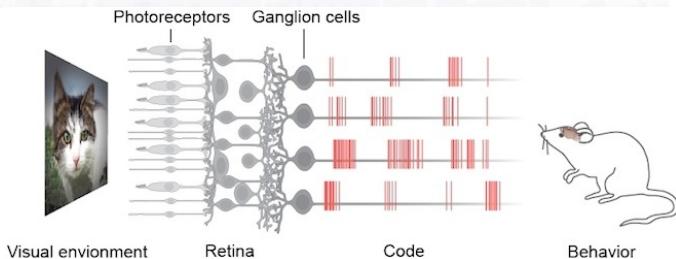
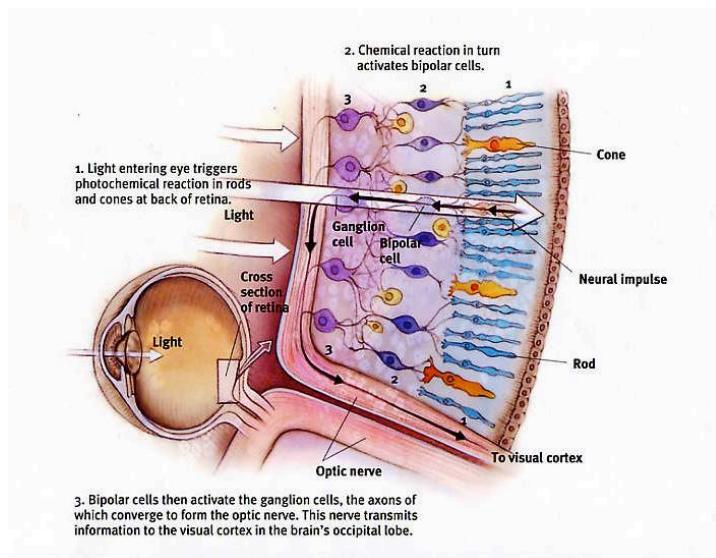
# Where we came from...



- Develop new **bidirectional methodology** to understand the brain
- Merging Computational and Biological Vision
- Importance of applications in Brain-Machine Interfaces

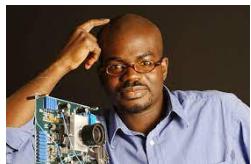
# Where we came from...

The retina: “the most approachable portion of the Brain”



# Where we came from...

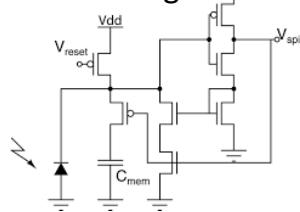
## Event-based Cameras: A Long Incremental Quest



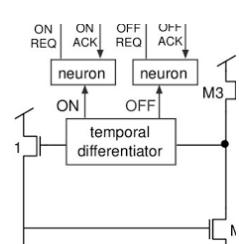
Kwabena Boahen



Ralph Etienne-Cummings



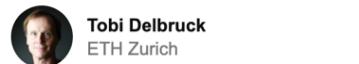
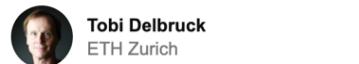
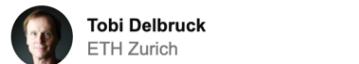
Jorg Kramer



Patrick Lichtsteiner  
ETH Zurich



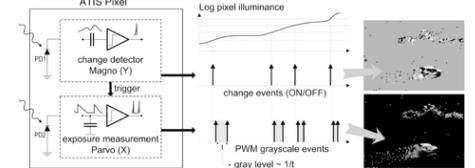
Tobi Delbrück  
ETH Zurich



Christoph Posch  
Sorbonne Université



Daniel Matolin



2000

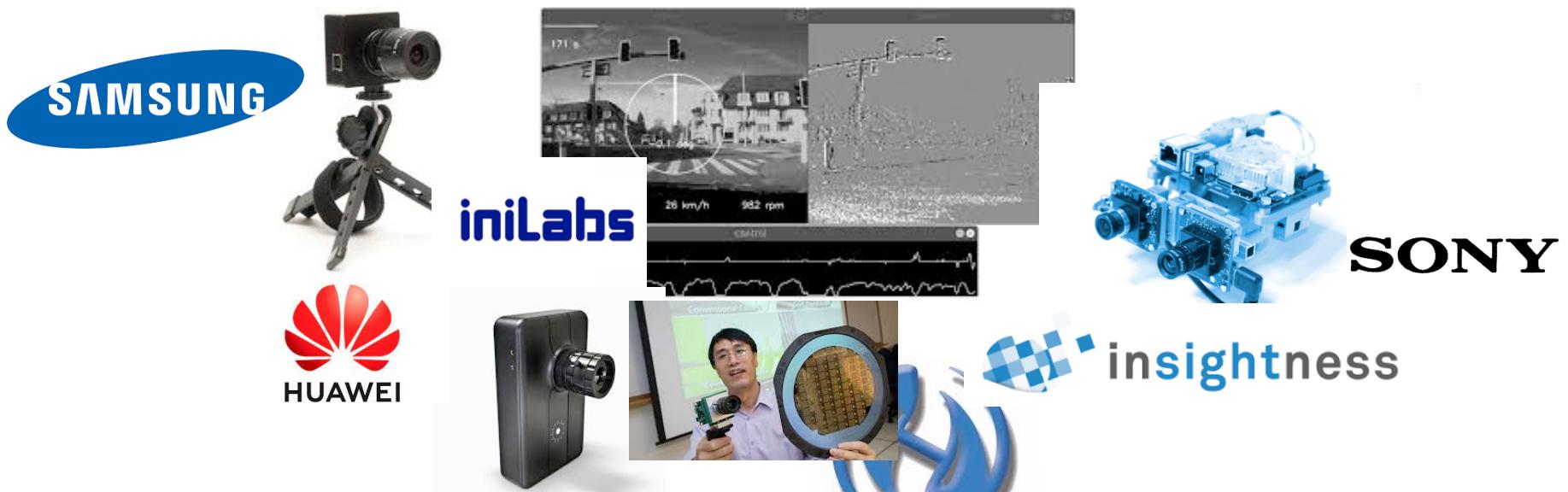
2004

2008



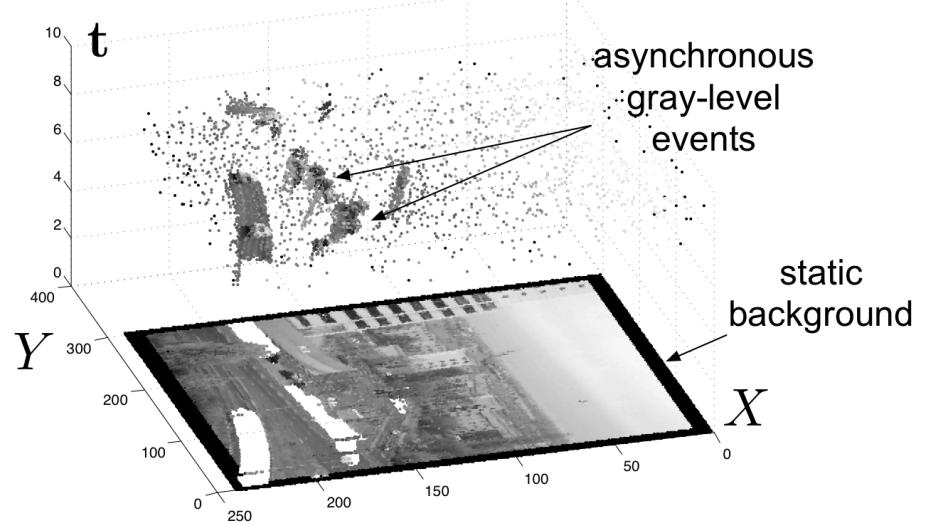
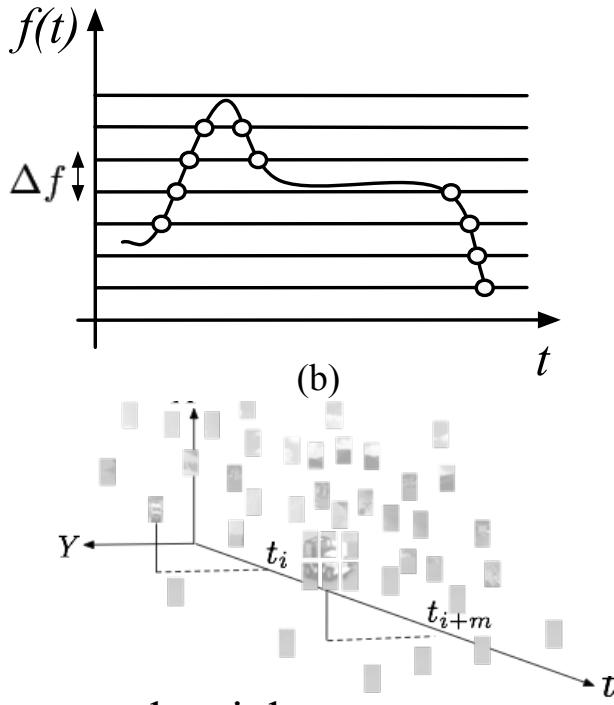
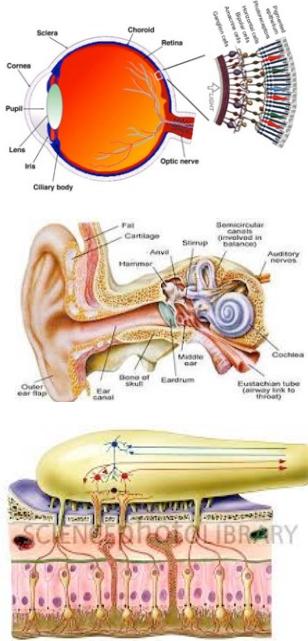
# Where we are ...

## Event-Based Cameras V1.0: Panorama



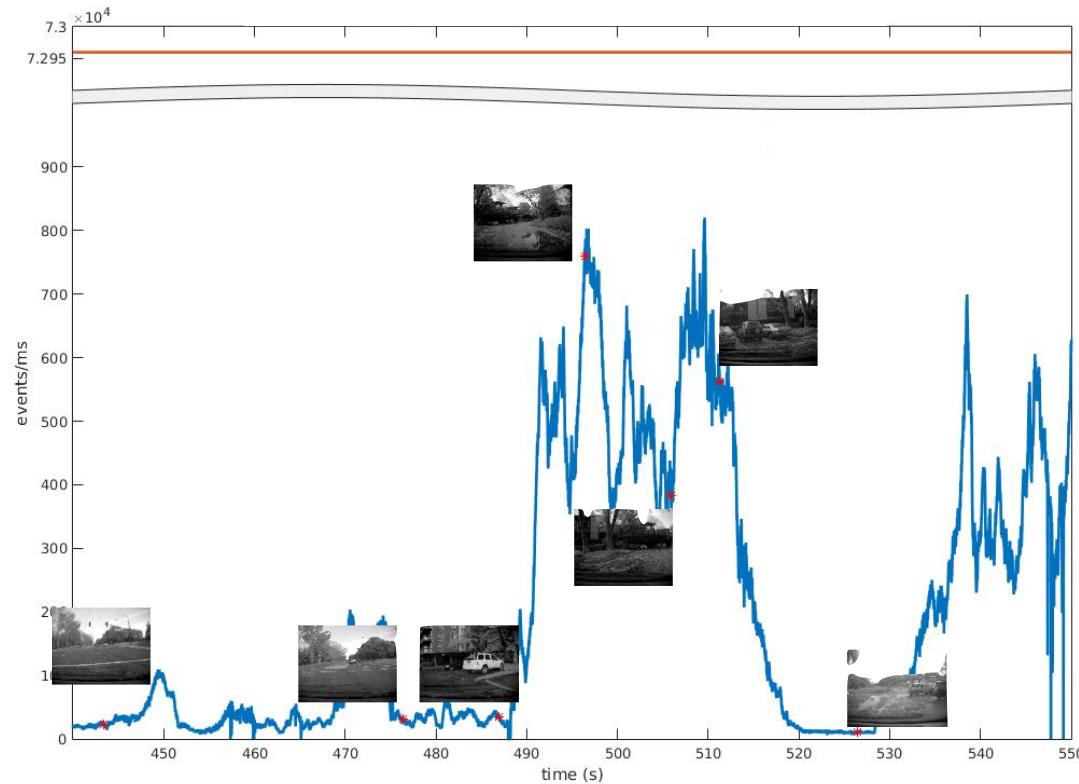
- Event-based cameras have become **a commodity**

# Event Acquisition



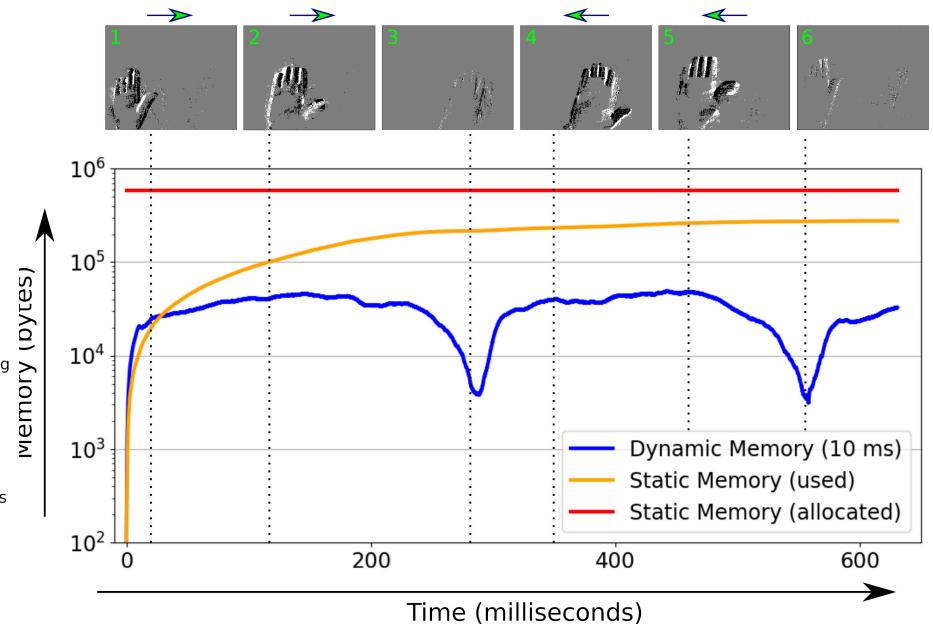
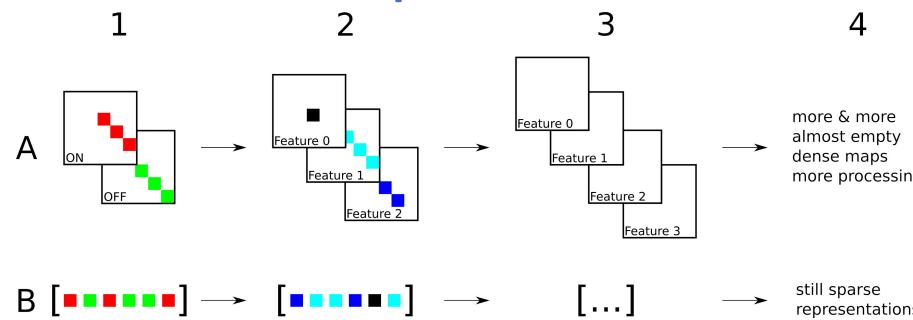
- Information is sent when it happens
- When nothing happens, nothing is sent or processed
- Sparse information coding
- Time is the most valuable information

# Properties Event Based sensors?



- Amount of Data generated from a moving car per 10 miliseconds

# Generating Frames is non optimal



Dataset (Mean Event Rate & Sensor Size)	PokerDVS 170.4 ev/ms 35x35			N-MNIST 13.6 ev/ms 28x28			DvsGesture 56.9 ev/ms 128x128			NavGesture-walk 188.6 ev/ms 304x240		
1. Time Window (ms)	1	10	100	1	10	100	1	10	100	1	10	100
2. Mean Number of events in TW (percentage of active pixels)	101 (8%)	390 (32%)	486 (40%)	22 (3%)	84 (11%)	229 (29%)	53 (<1%)	340 (2%)	1751 (11%)	285 (<1%)	2818 (4%)	13279 (18%)
3. Max Number of events in TW (percentage of active pixels)	356 (29%)	848 (69%)	1052 (86%)	223 (28%)	312 (40%)	597 (76%)	467 (3%)	2056 (13%)	9191 (56%)	2599 (4%)	18296 (25%)	68128 (93%)
4. Working Memory Size (kB) Dynamic - Average case	0.8	3.1	3.9	0.2	0.7	1.8	0.4	2.7	14.0	2.3	22.5	106.2
5. Working Memory Size (kB) Dynamic - Worst case	2.8	6.8	8.4	1.8	2.5	4.8	3.7	16.4	73.5	20.8	146.3	545.0
6. Allocated Memory Size (kB)	9.8	9.8	9.8	6.3	6.3	6.3	131	131	131	584	584	584
7. Memory ratio dynamic/static (Average Case)	8%	32%	40%	3%	11%	29%	1%	2%	11%	1%	4%	18%
8. Memory ratio dynamic/static (Worst Case)	29%	69%	86%	28%	40%	76%	3%	13%	56%	4%	25%	93%

# Computer vision: the impossible trade off of power vs frame rate



Too many data

Too slow

Light-dependent

## over-sampling

- ▶ redundant useless data
- ▶ power and resource hungry: need to acquire/transmit/store/process

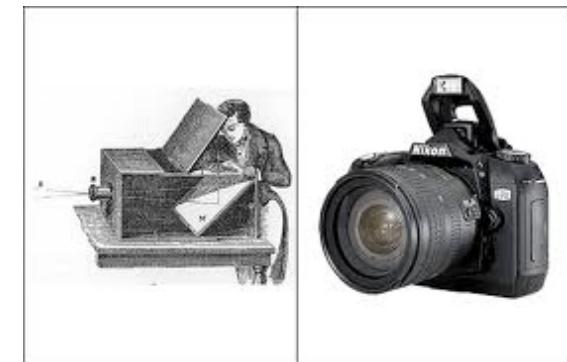
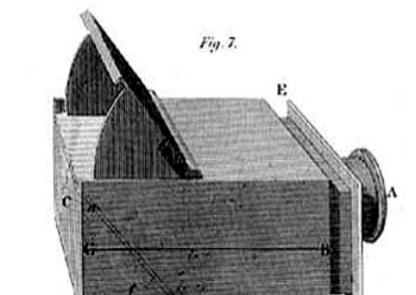
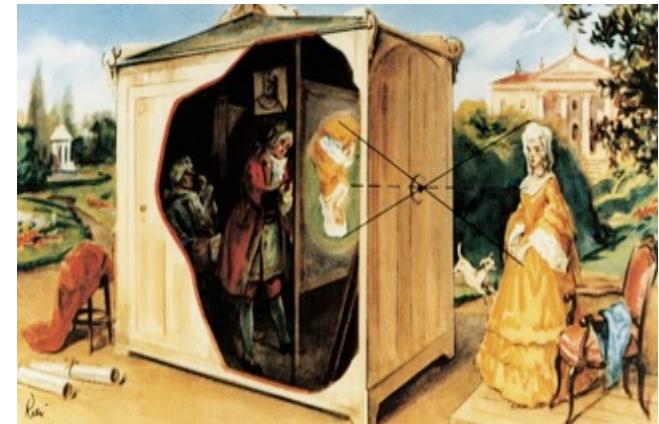
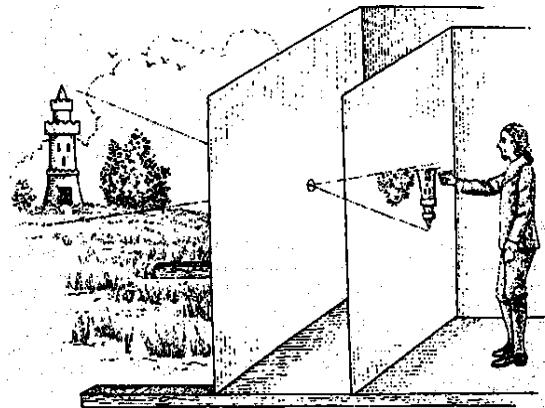
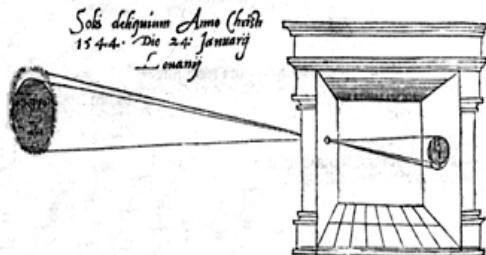
## under-sampling

- ▶ motion blur
- ▶ displacement between frames

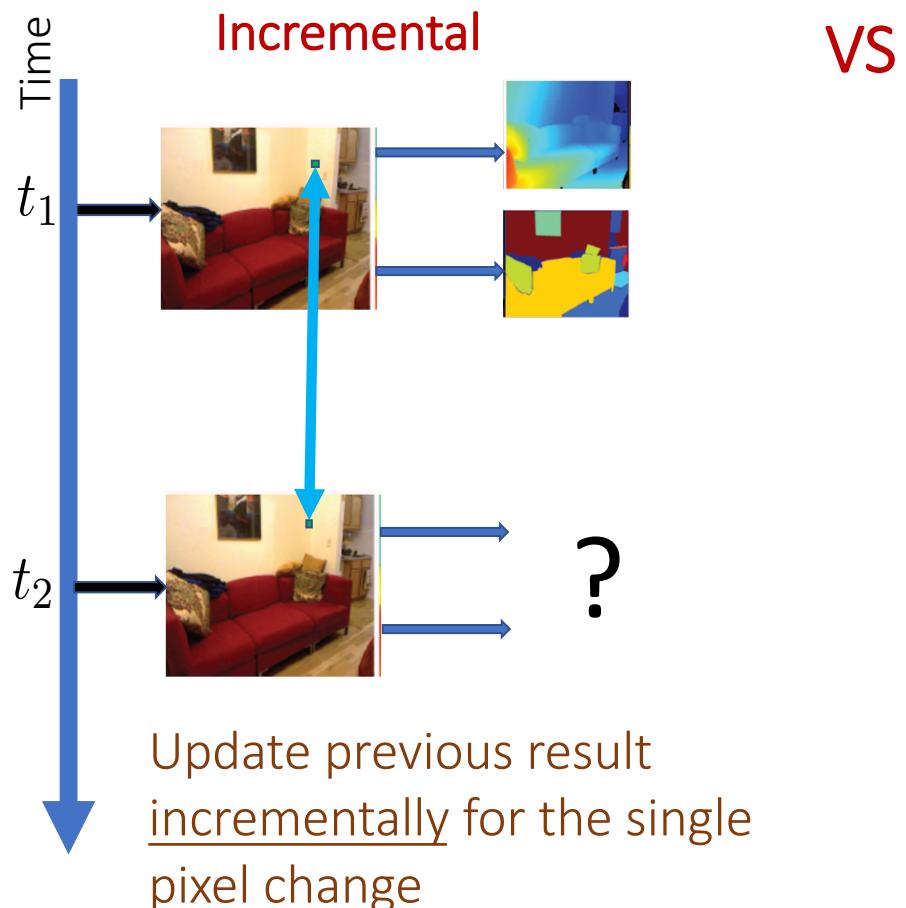
High Power & High Latency

# Origins of Imaging

illum in tabula per radios Solis, quam in celo contin-  
git: hoc est, si in celo superior pars deliqui patiatur, in  
radii apparet in inferior deficere, vt ratio exigit optica.

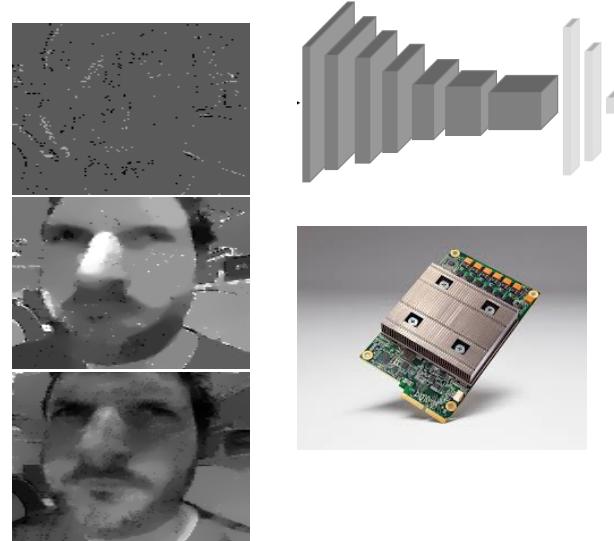


## Event Computation



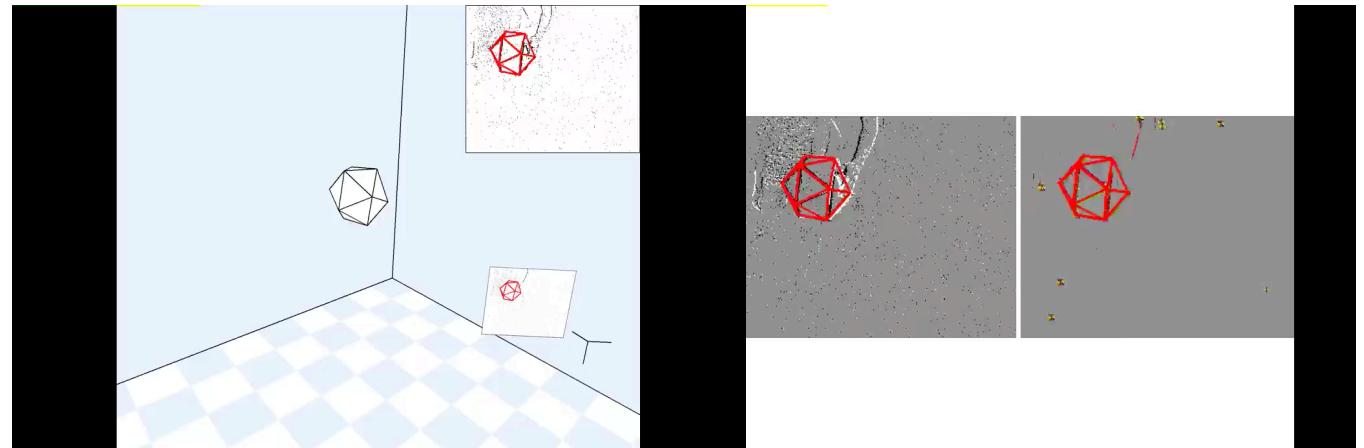
VS

Batch or Frame

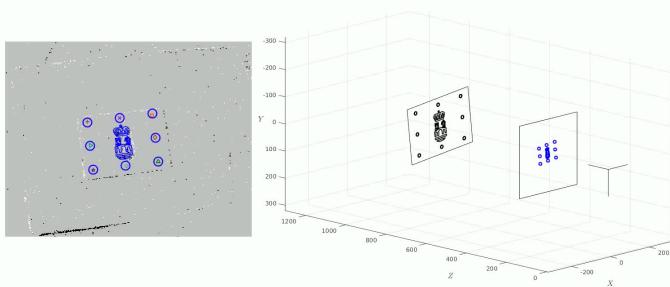


Compute the new 1 pixel-change frame and compute result using the whole frame

# Incremental Event Vision: everything can be written as an incremental event process



Event-Based Solution to the PnP Problem



Event-based Face Detection in the Blink of an Eye

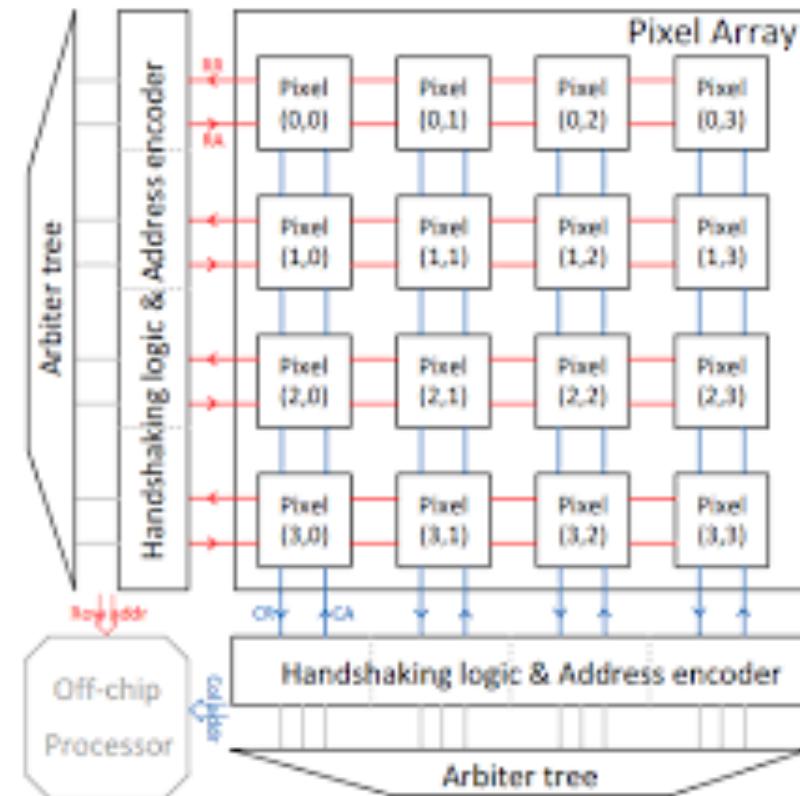


# Where we are heading to



## Replace arbitrers

- Current locks are caused by the arbiter that scrambles event times and makes computation difficult
- Event Cameras are becoming compressed imagers.

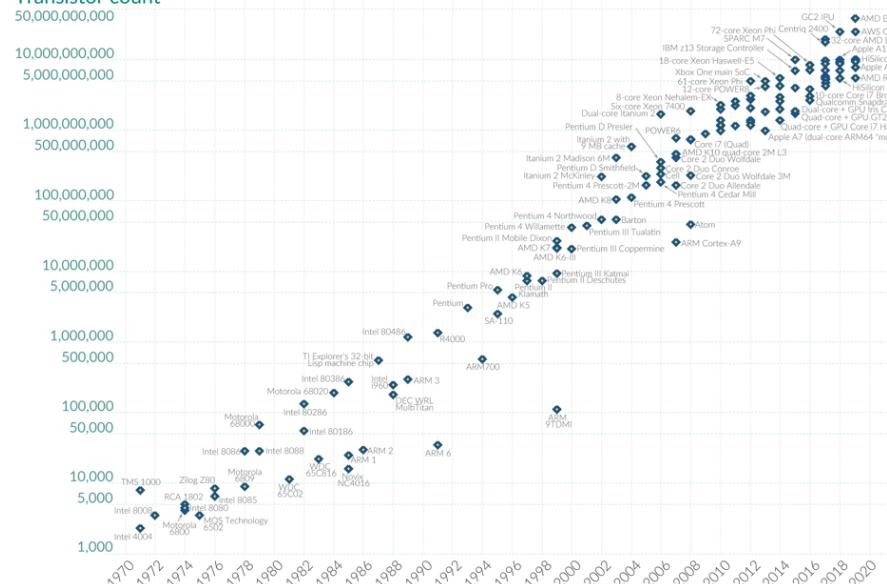


# Where we are heading to: the promise

Moore's Law: The number of transistors on microchips doubles every two years

Moore's law describes the empirical regularity that the number of transistors on integrated circuits doubles approximately every two years. This advancement is important for other aspects of technological progress in computing – such as processing speed or the price of computers.

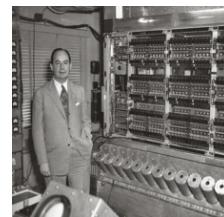
## Transistor count



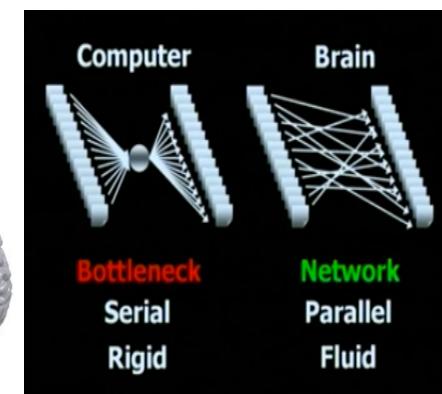
Data source: Wikipedia ([wikipedia.org/wiki/Transistor\\_count](https://en.wikipedia.org/w/index.php?title=Transistor_count&oldid=1000000000)) Year in which the microchip was first introduced

[OurWorldinData.org](#) – Research and data to make progress against the world's largest problems.

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## Von Neumann bottleneck

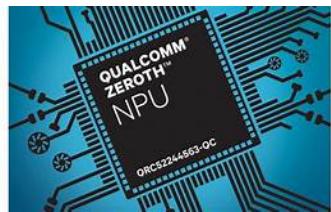


# Plateau in current main stream computing technologies

Promise of a new path solving for:  
**Low Power and Latency**

# Where we are heading to

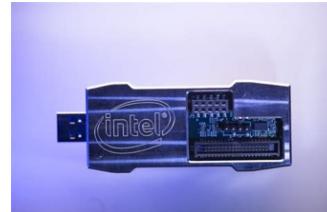
Existing Neuromorphic Processing Hardware is based on what we know of the brain



Qualcomm Zeroth (2013)



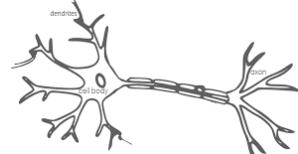
IBM TrueNorth (2014)



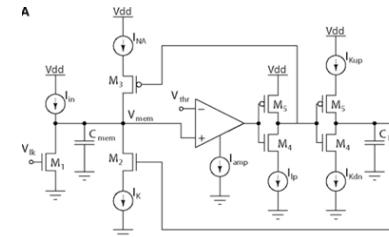
Intel Loihi (2017)



BrainChip (2019)



replicate



Existing hardware is based on the concept of replicating biological neurons into silicon  
Limited use cases!

# Where we are heading to

Replicating nature's solutions is not always the optimal path to solve an engineering problem.



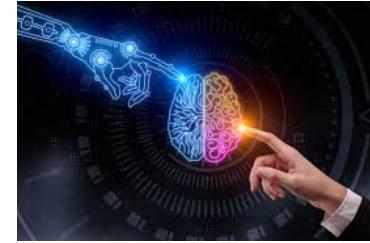
understand



Understanding rather than replicating

There is a need to **find the right level of abstraction**

# Where We Are Heading To



- We should explore new forms of events' acquisition
- Find a better link between images and events and see how to connect with decades of CV without losing the advantages of events
- We need a dedicated processor adapted to the temporal precision and sparseness of data of events and the amount of generated data
- Event cameras are the future if we explore their temporal properties
- New kind of engineers that understand neurosciences where « biological » events are studied