

Event-Driven Sensing for Autonomous Robots

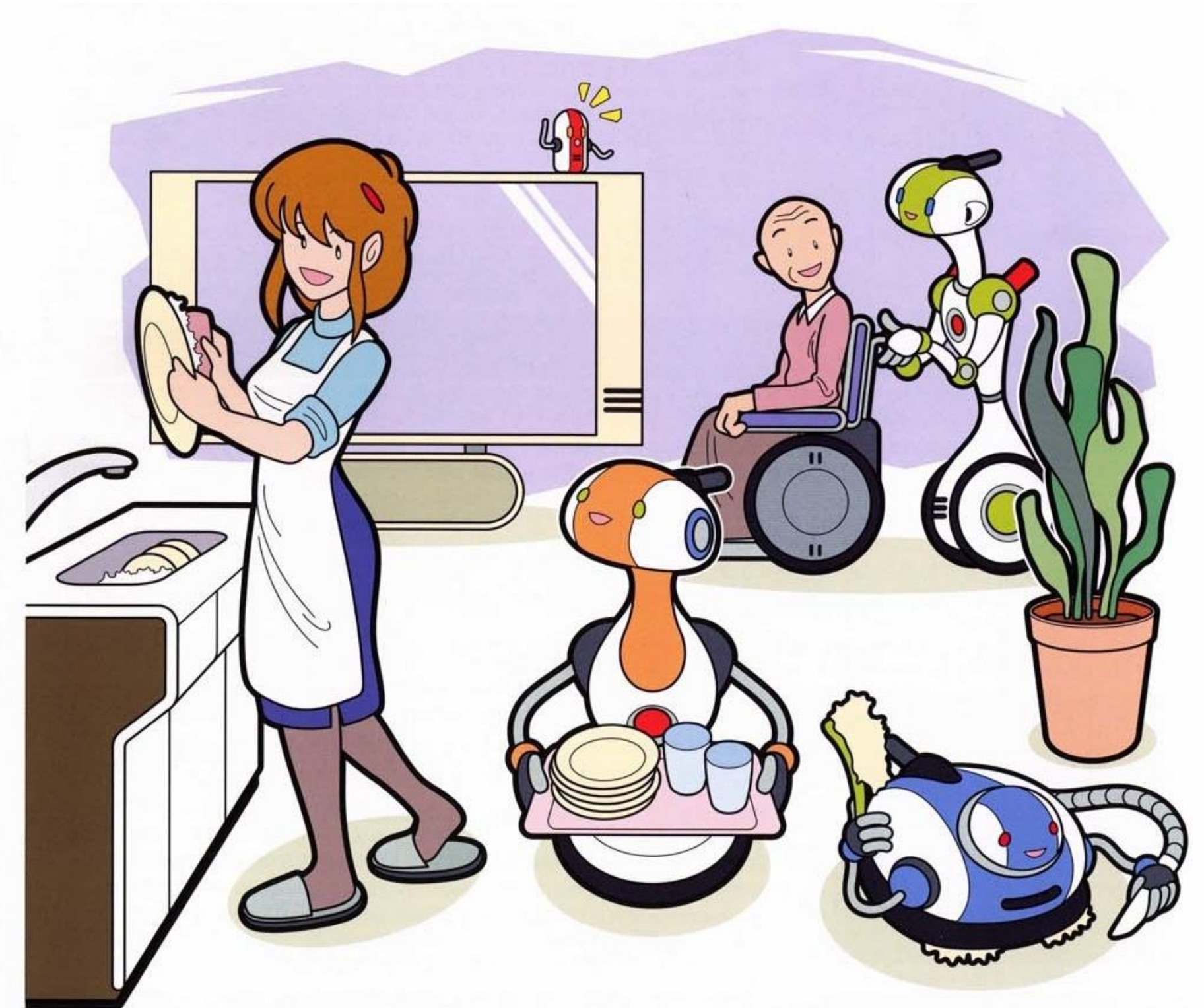
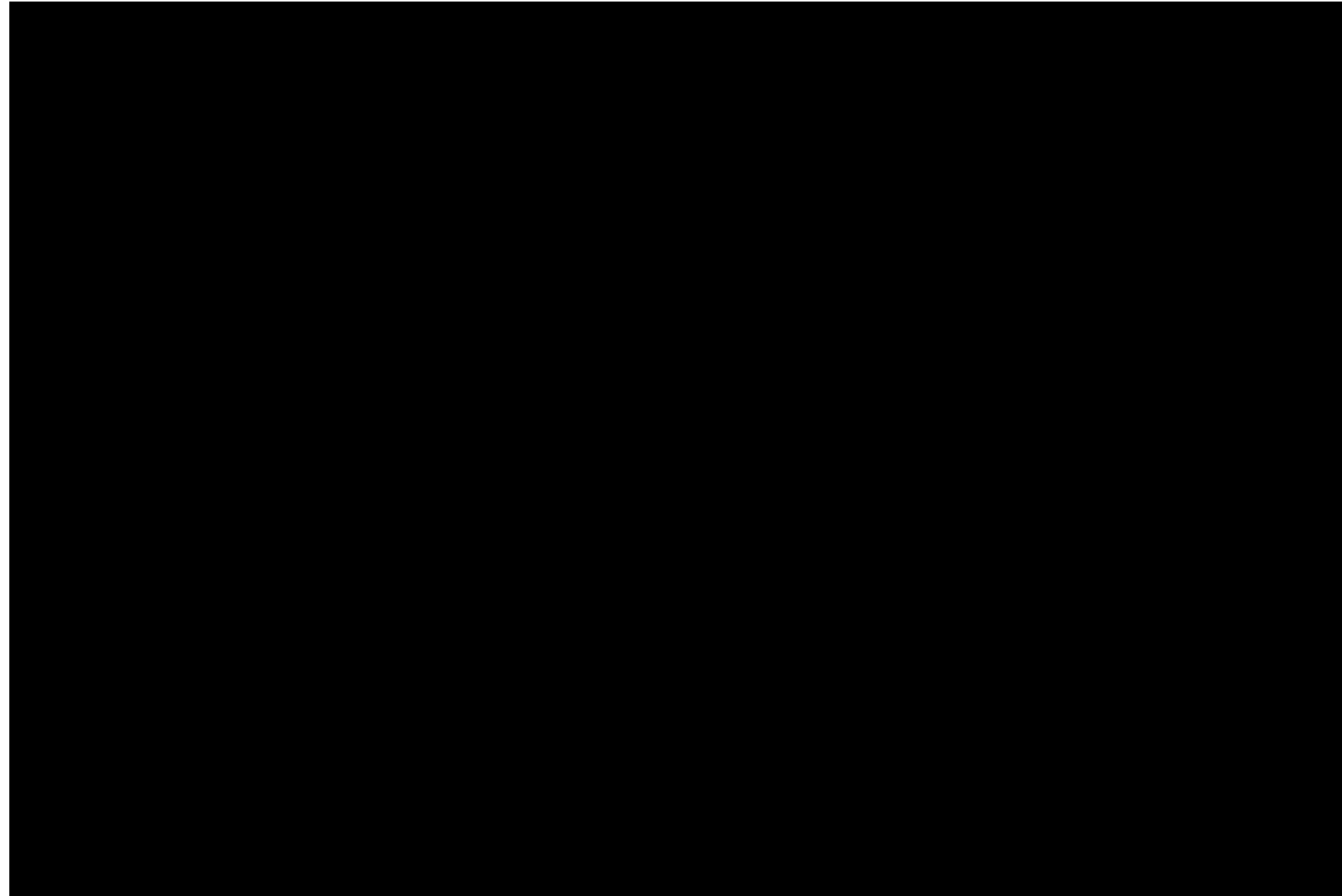
Arren Glover

Senior Post-doc | Italian Institute of Technology



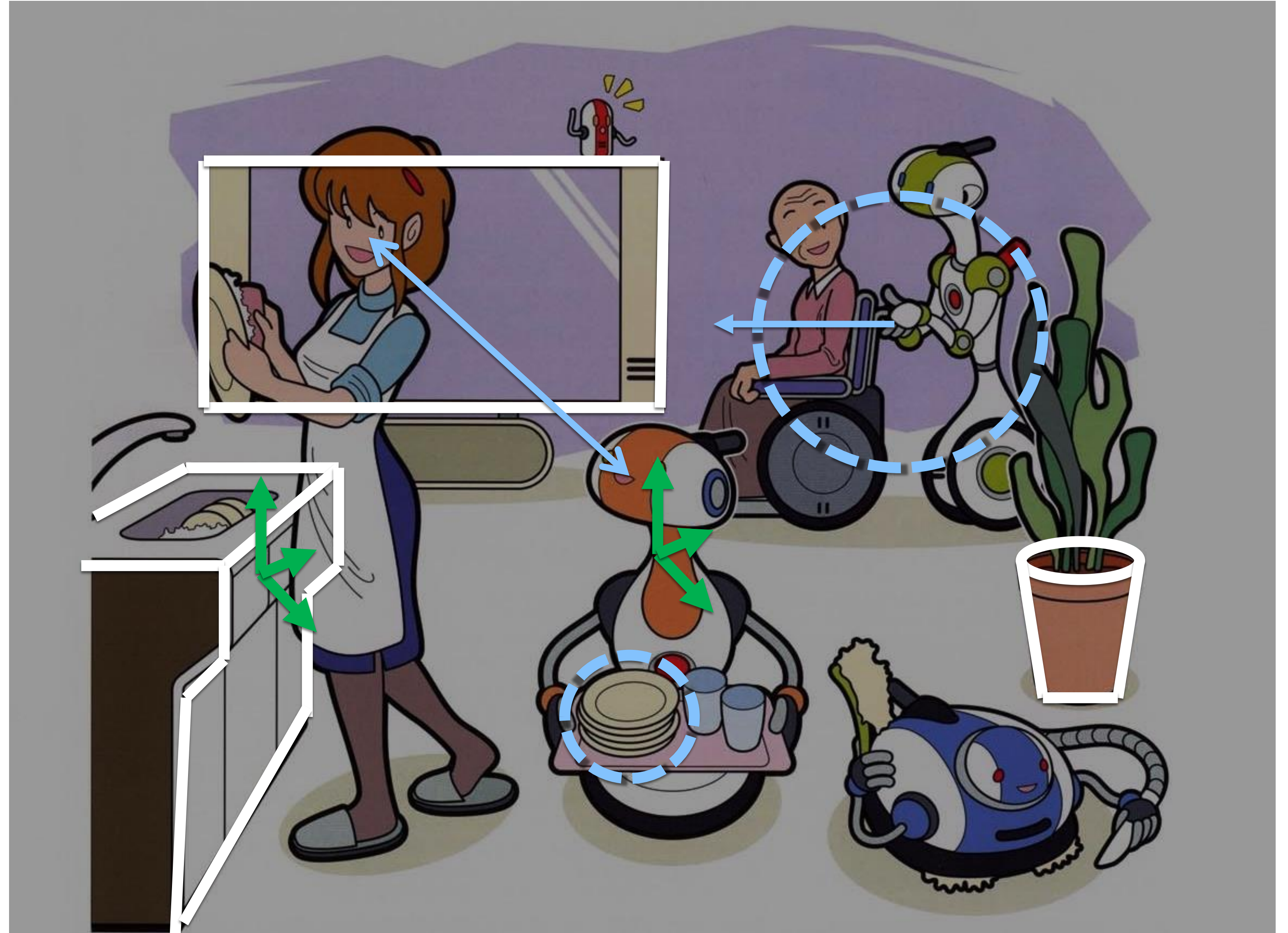
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The iCub Humanoid Robot



What is required for autonomy?

- Model of environment
- Robot pose relative to environment
- Tasks requires searching for objects
- Track dynamic objects (avoiding/interacting)
- Interaction with humans (learning)



Roadblocks to autonomy?

Sensors

- Aren't suitable for the task
- Don't work under all conditions

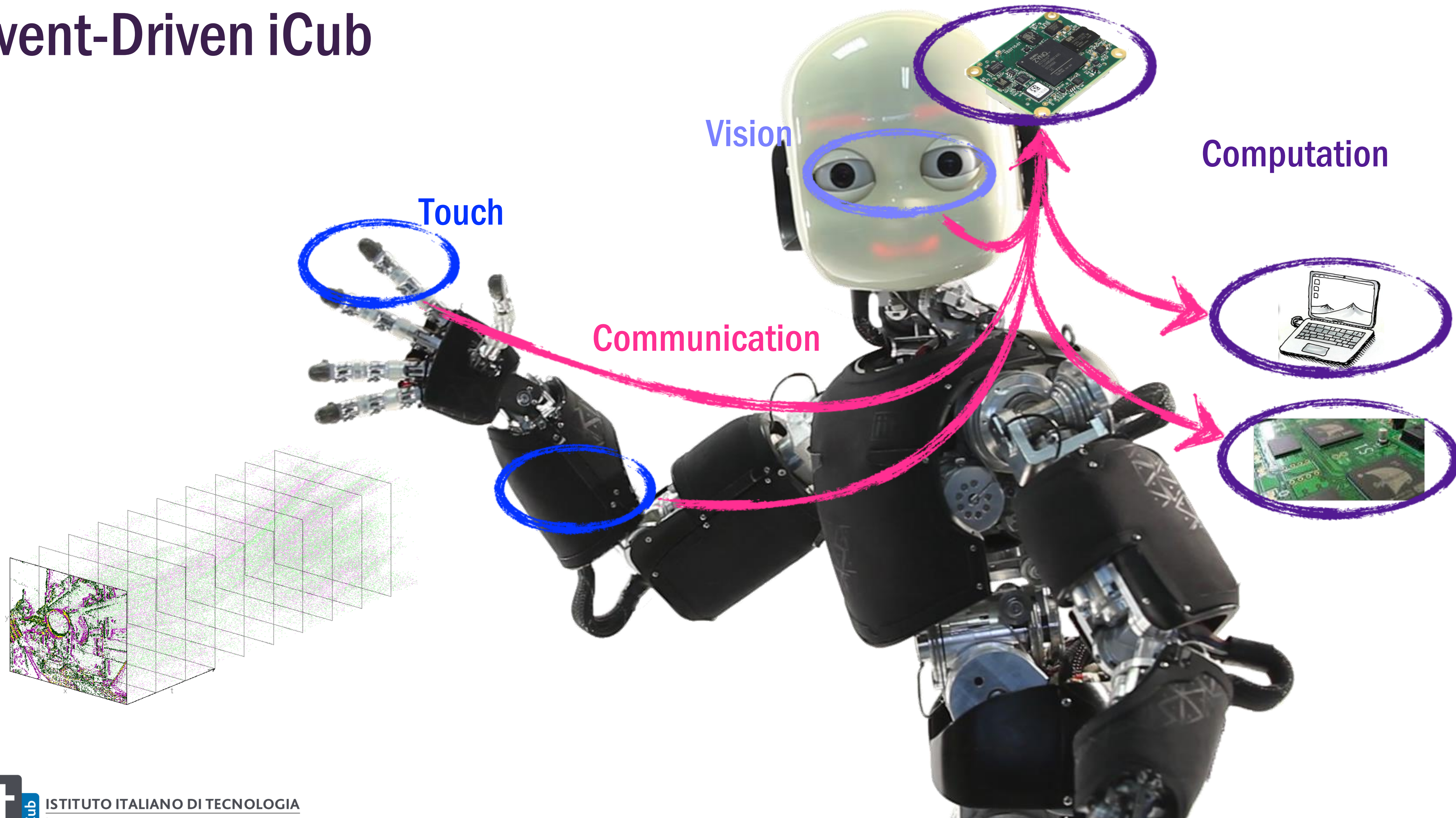
Processing

- Shouldn't rely on wireless communications for safety critical systems
- On-board systems need to be low resource consumers

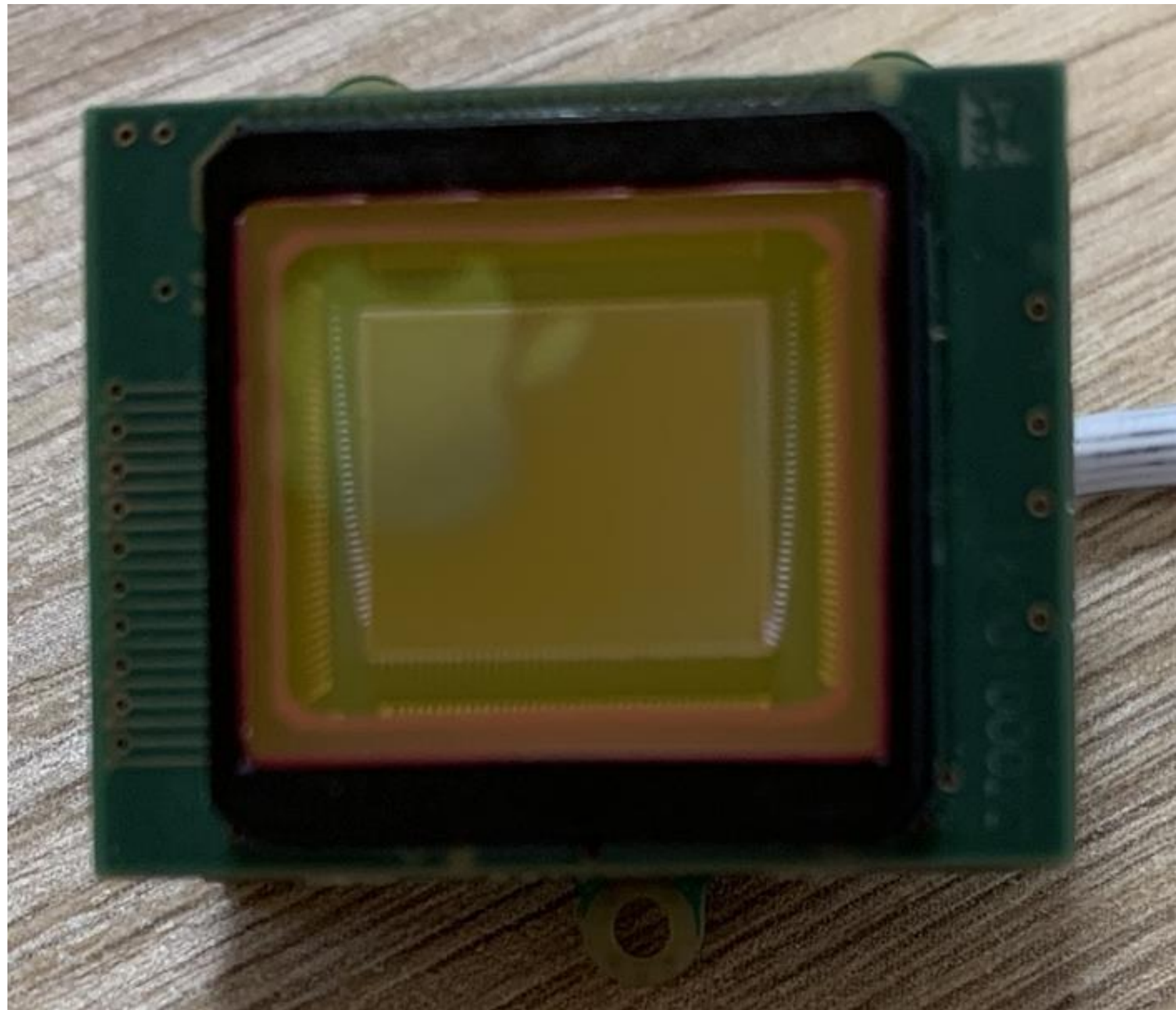
System Integration

What role does neuromorphic engineering play?

Event-Driven iCub



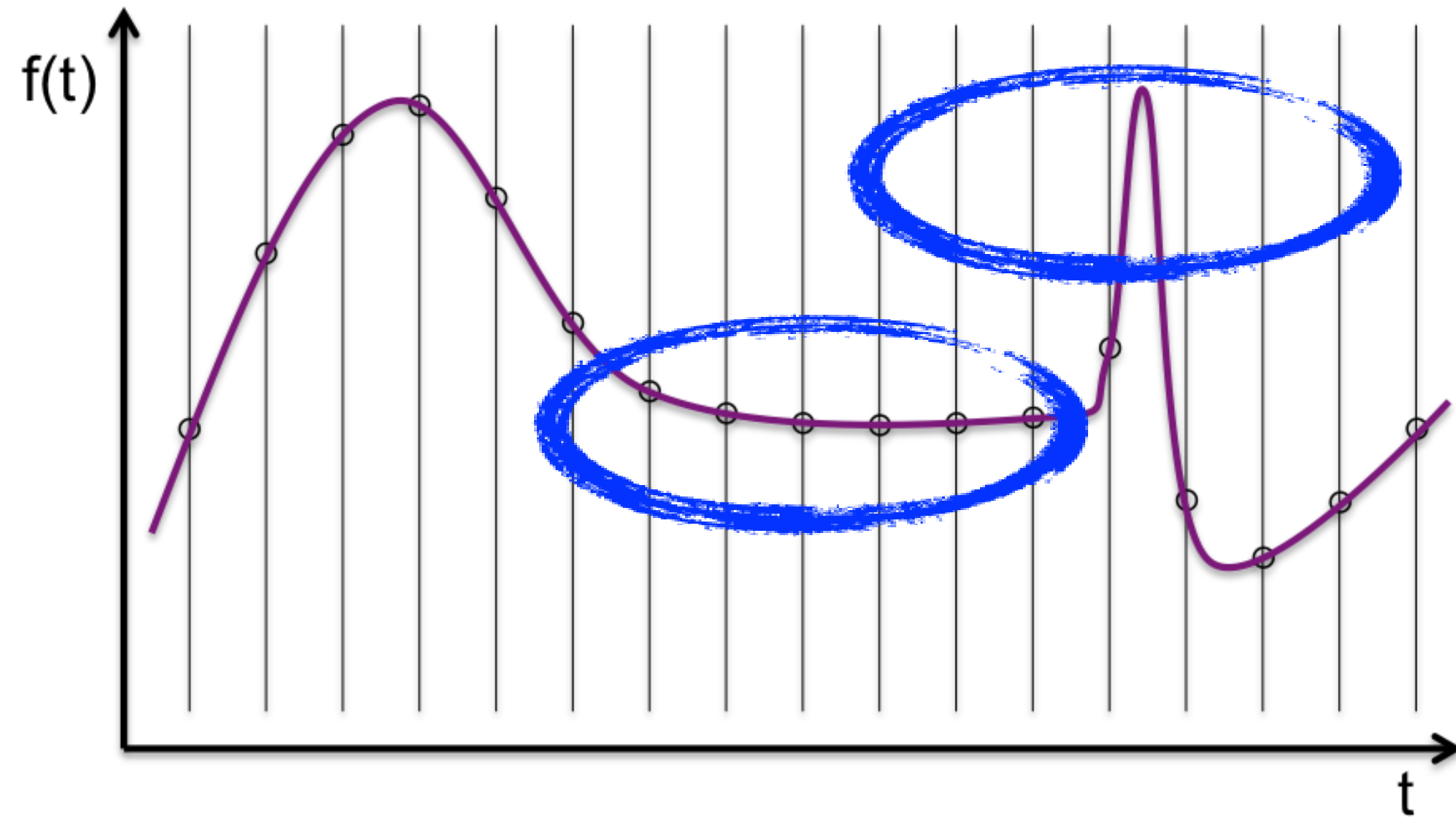
Event cameras (ATIS)



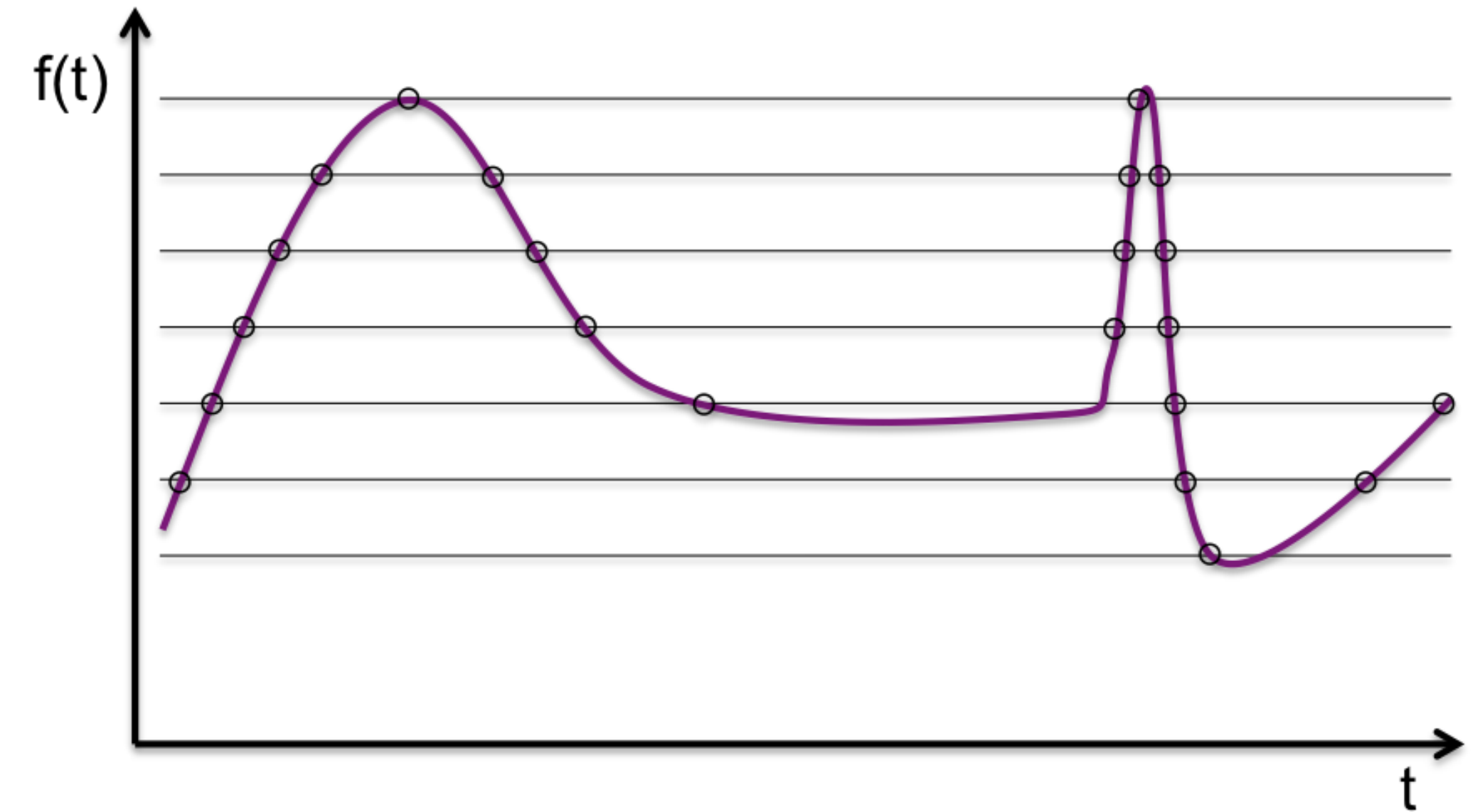
- Inspired by biological vision
- Each pixel has an independent circuit
- Measures change in light
- Produces an output when change occurs

Event-Driven Sensing and Processing

Clock-Based Sampling – fixed Δt

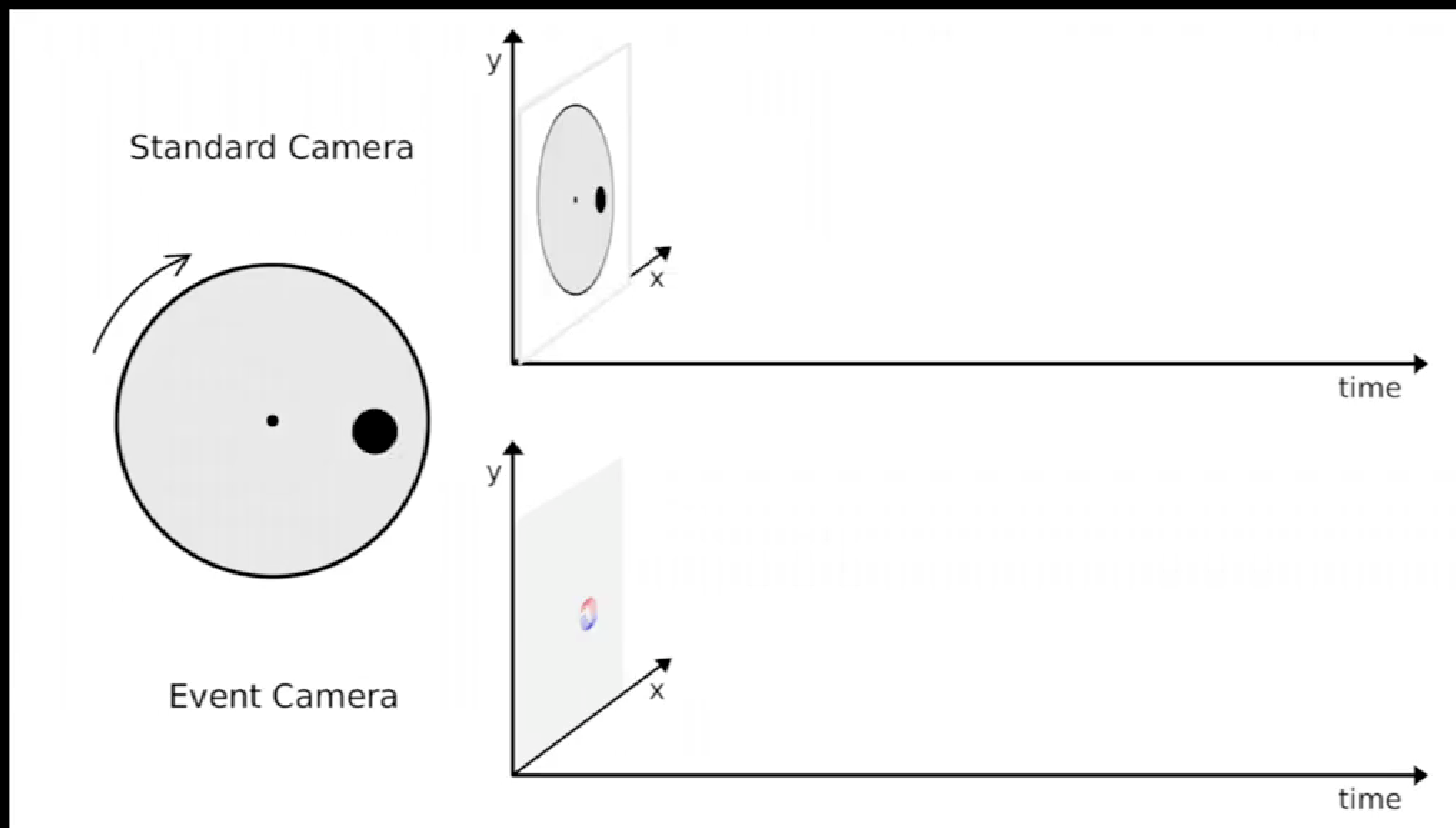


Data-Driven Sampling – fixed Δf



Event-Driven Vision [1]

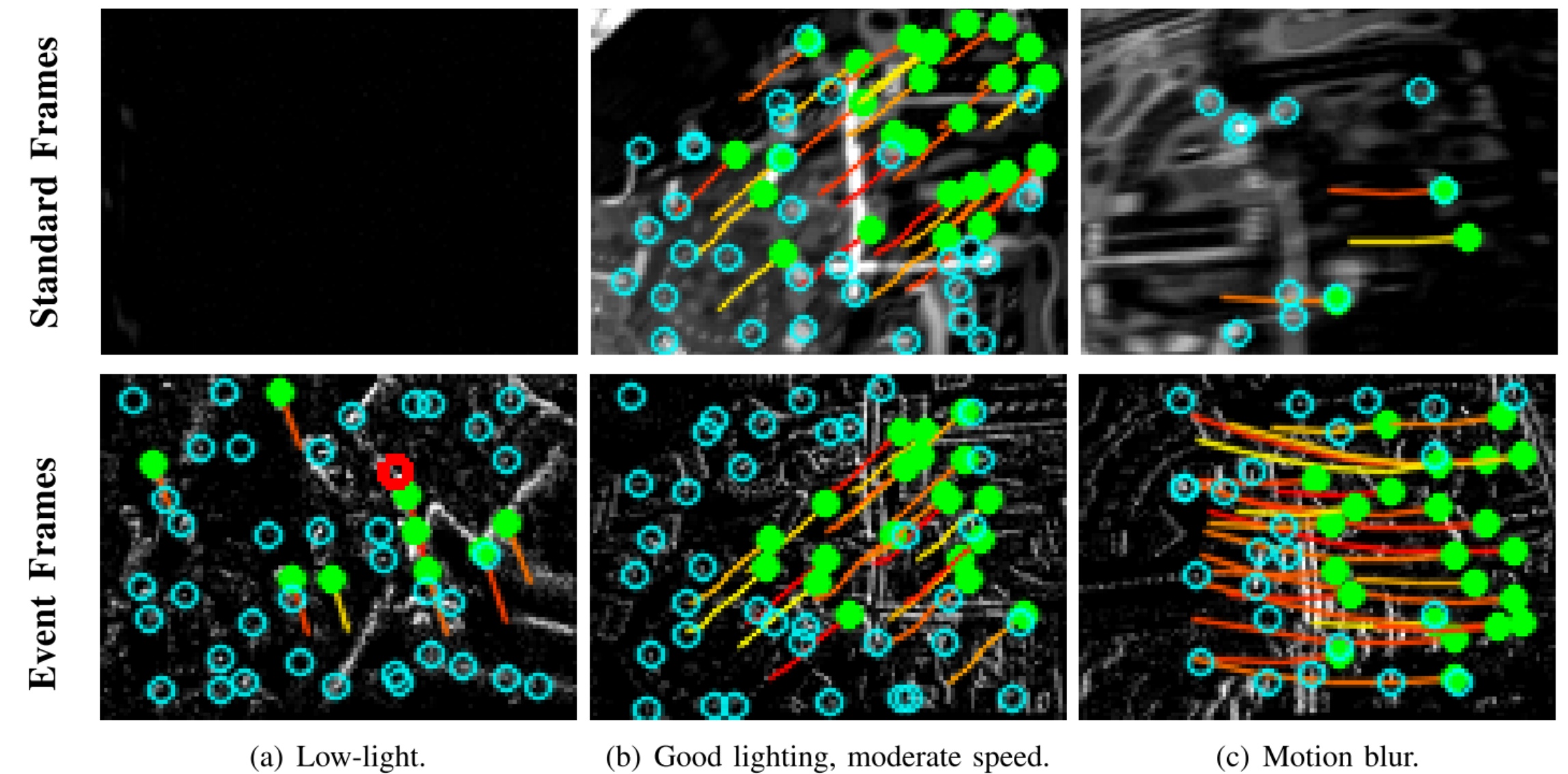
only informative pixels are provided



We recreated this animation inspired by the associated animation of Mueggler, Huber, and Scaramuzza, "Event-based, 6-DOF Pose Tracking for High-Speed Maneuvers", IROS, 2014: <https://youtu.be/LauQ6LWTkxM?t=35s>.

Event-cameras for Robotics

- High dynamic range
- Temporal precision
 - Understand what happens “between frames”
 - Avoid motion-blur
- Signal compression
 - 0 mbps with no motion
 - 10 – 120 mbps normal/heavy conditions
 - > 300 mbps frame-based cameras
- Integration with spiking networks / neural systems



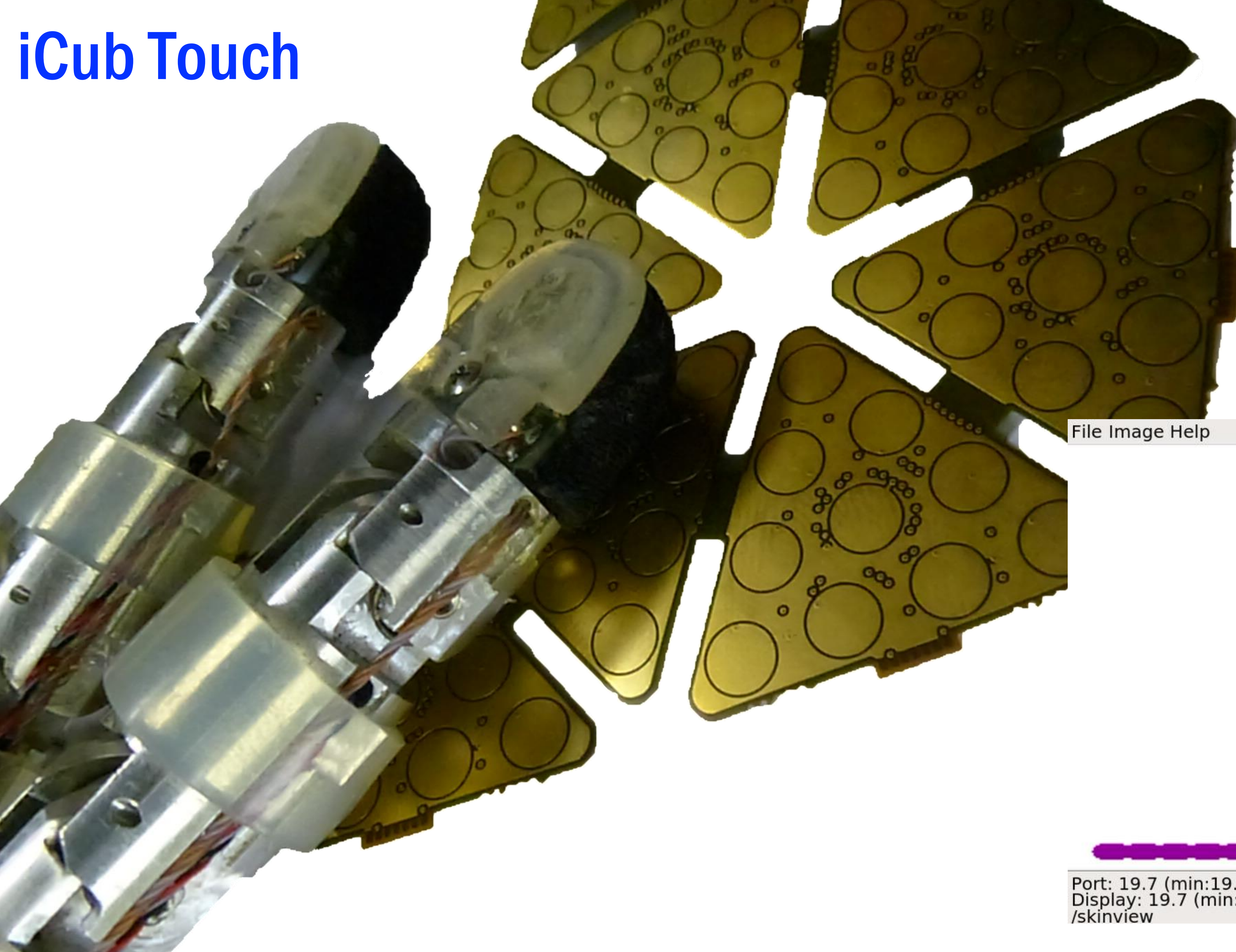
(a) Low-light.

(b) Good lighting, moderate speed.

(c) Motion blur.

[1] A. R. Vidal, H. Rebecq, T. Horstschafer, and D. Scaramuzza, “Ultimate SLAM? Combining Events, Images, and IMU for Robust Visual SLAM in HDR and High Speed Scenarios,” *IEEE Robot. Autom. Lett.*, 2018.

iCub Touch

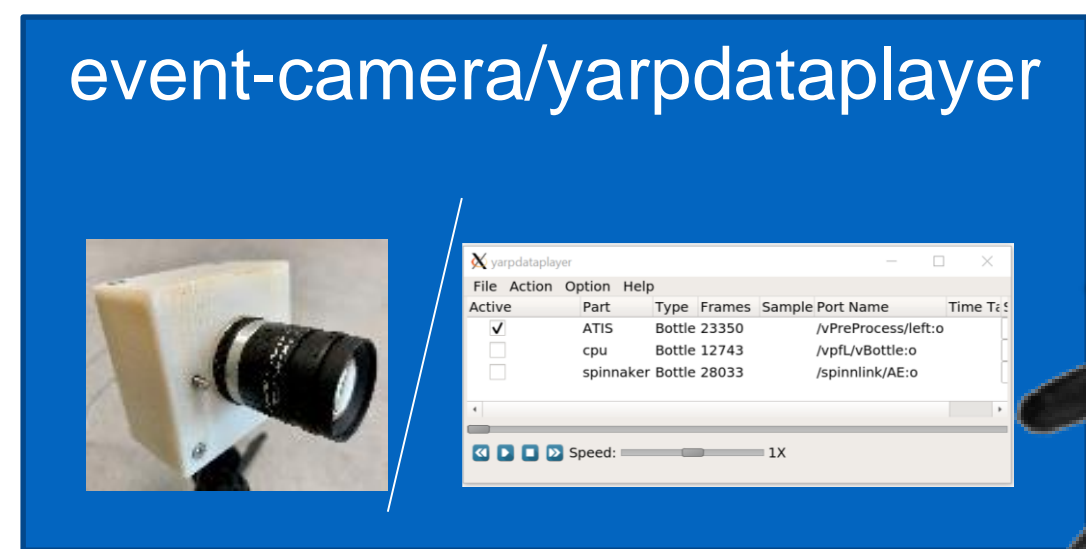


File Image Help

Port: 19.7 (min:19.2 max:20.5) fps
Display: 19.7 (min:19.2 max:20.5) fps
/skinview

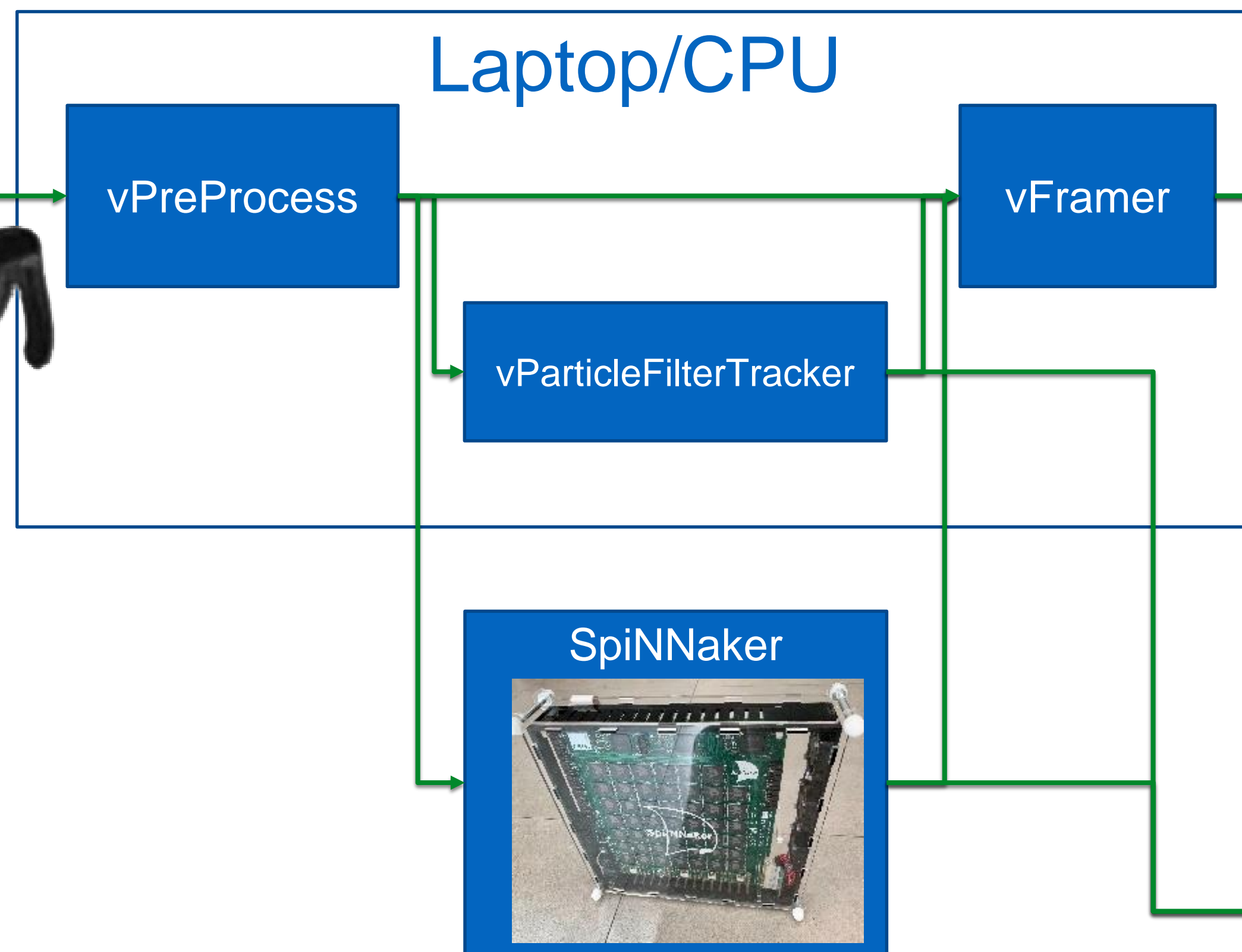
YARP and event-driven-libraries

event-camera/yarpdataplayer



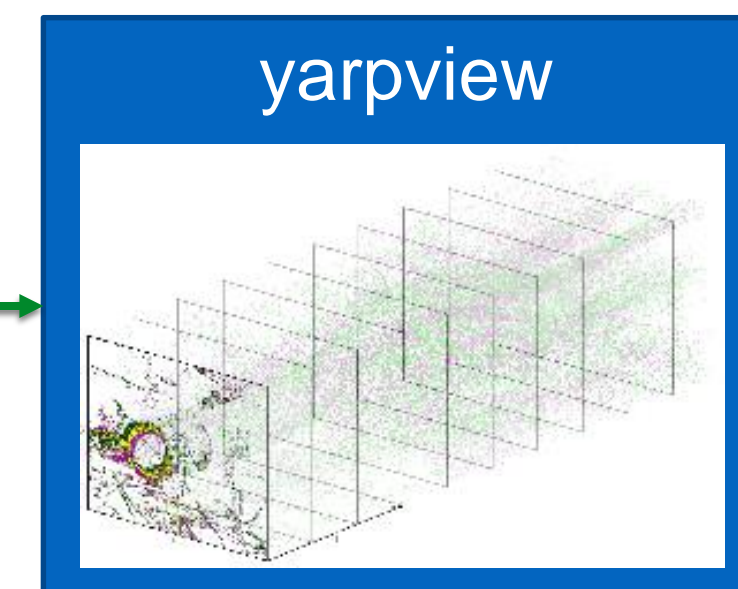
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<input type="checkbox"/>	spinnaker	Bottle	28033	/spinnlink/AE:o	

YARP Ports
(spikes)



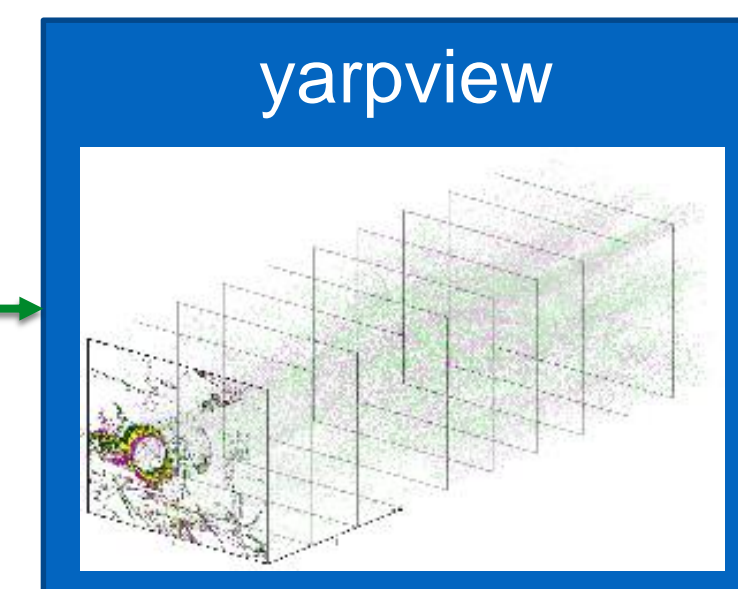
Fixed-rate

yarpview



Frame
refresh-rate
60 Hz

yarpview



action control

Motor
control loop
100 Hz



Spiking using event-driven-libraries (open source)

Hybrid approach to neural robotics

Take advantage of Neuromorphic Hardware

Take advantage of decades of robotics development

Choose the most appropriate algorithm for the task

Difficult to go from 0 to 100% completely neural for a complex robot. Modularity of robot middleware allows us to start with a “working” robot and swap in neuromorphic modules

What is required for autonomy?

- Track dynamic objects (avoiding/interacting)



Using event-cameras on a robot

Moving object with a
stationary ED camera



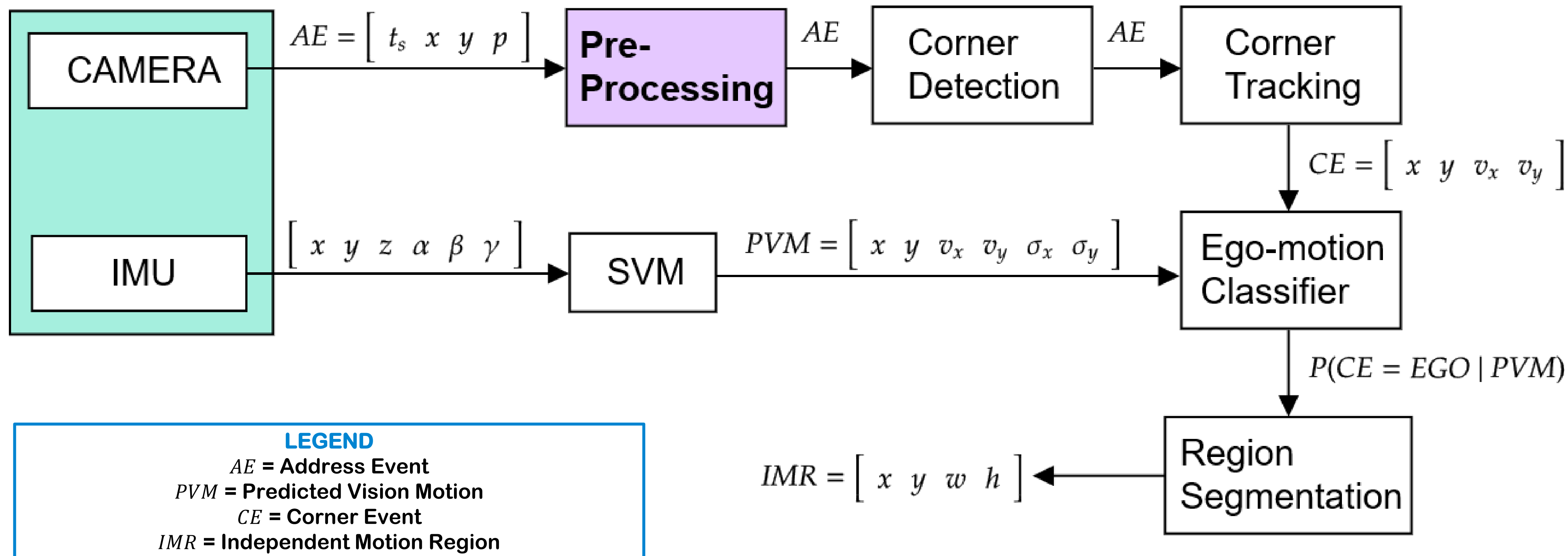
Moving object with a
moving ED camera



Vasco. V., Glover. A., Mueggler. E., Scaramuzza. D., Natale. L., and Bartolozzi. C.: Independent Motion Detection with Event-driven Cameras. In the Proceedings of the International Conference on Advanced Robotics (ICAR) 2017

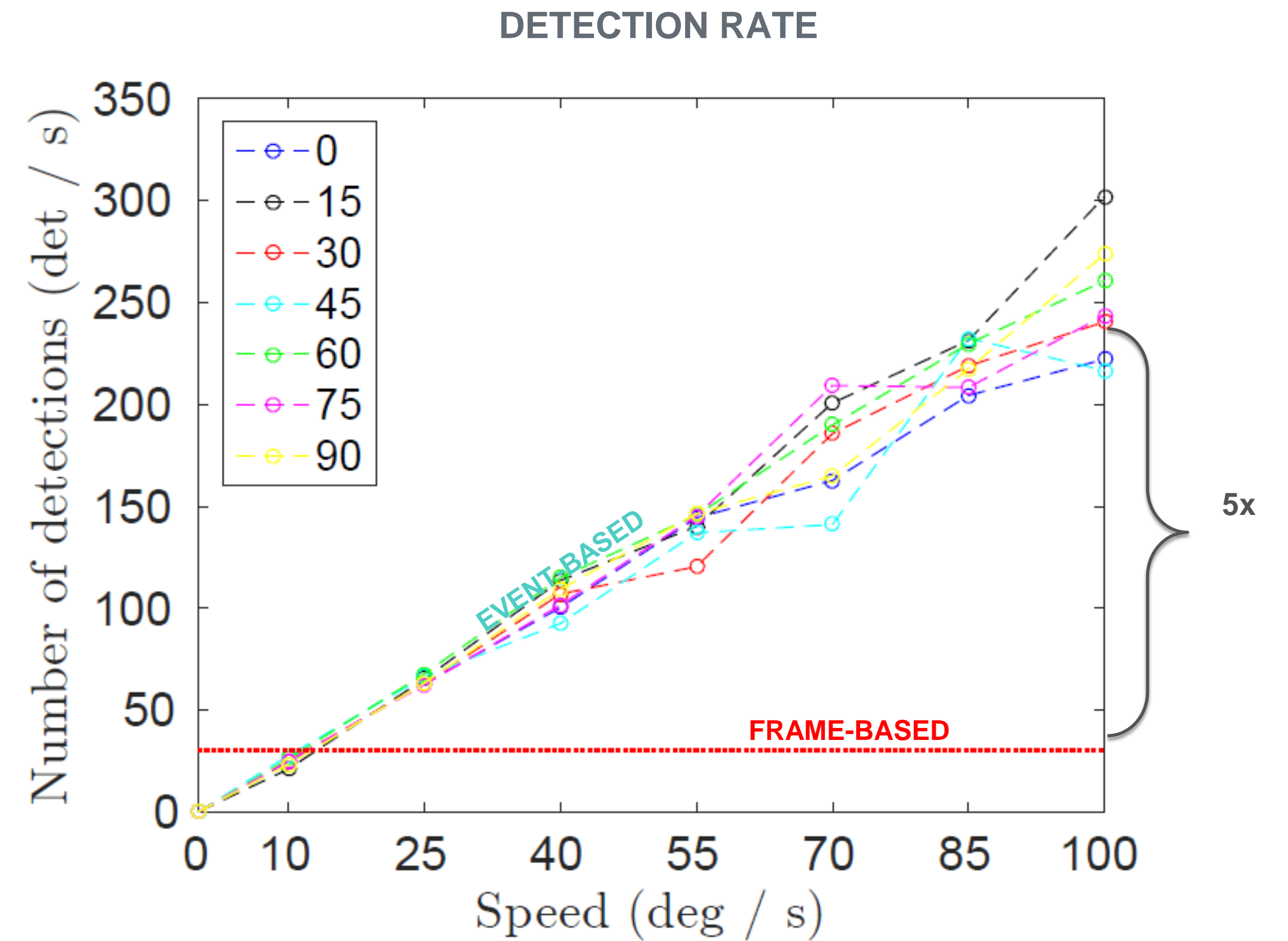
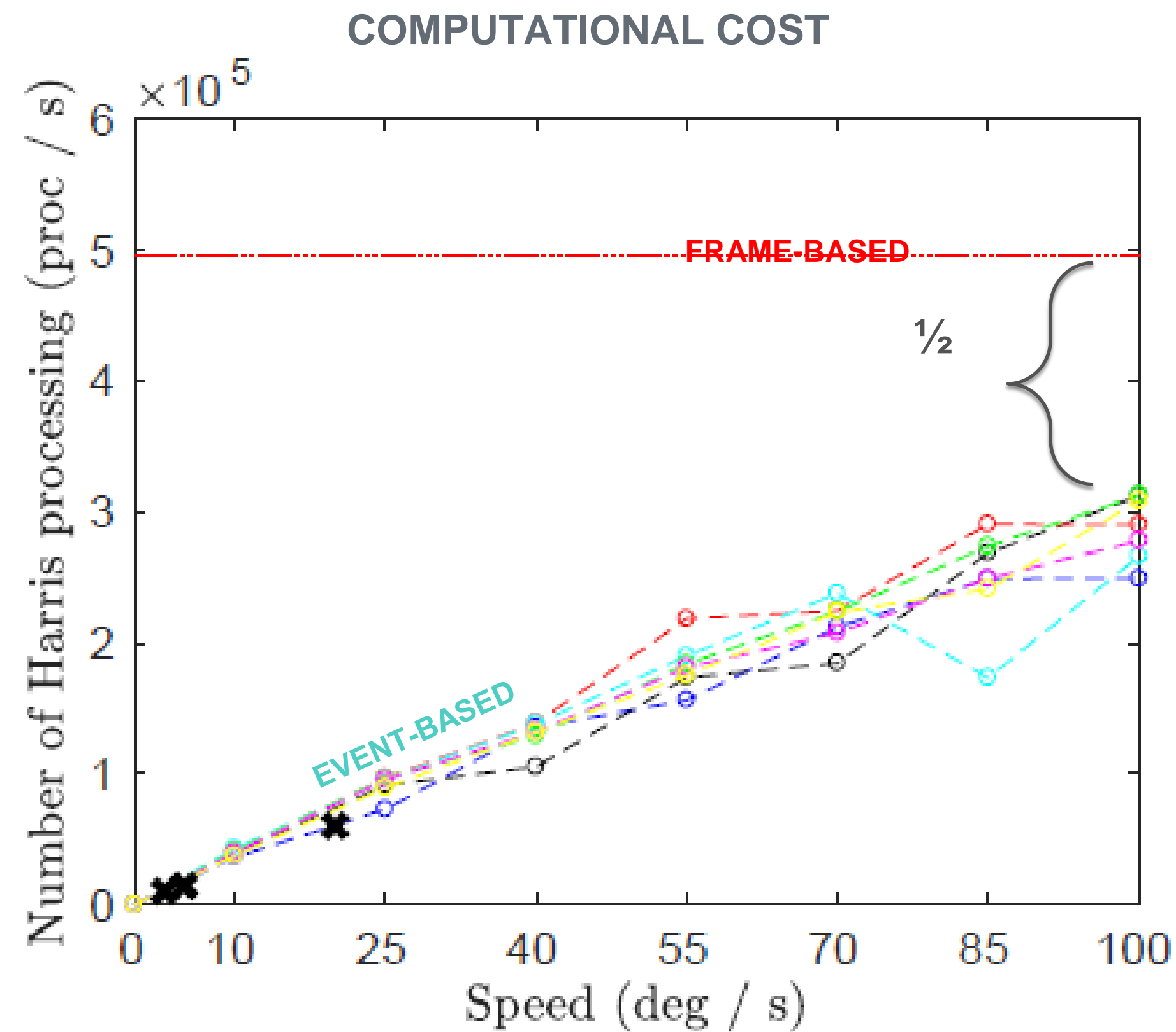
Detection of Independent Motion

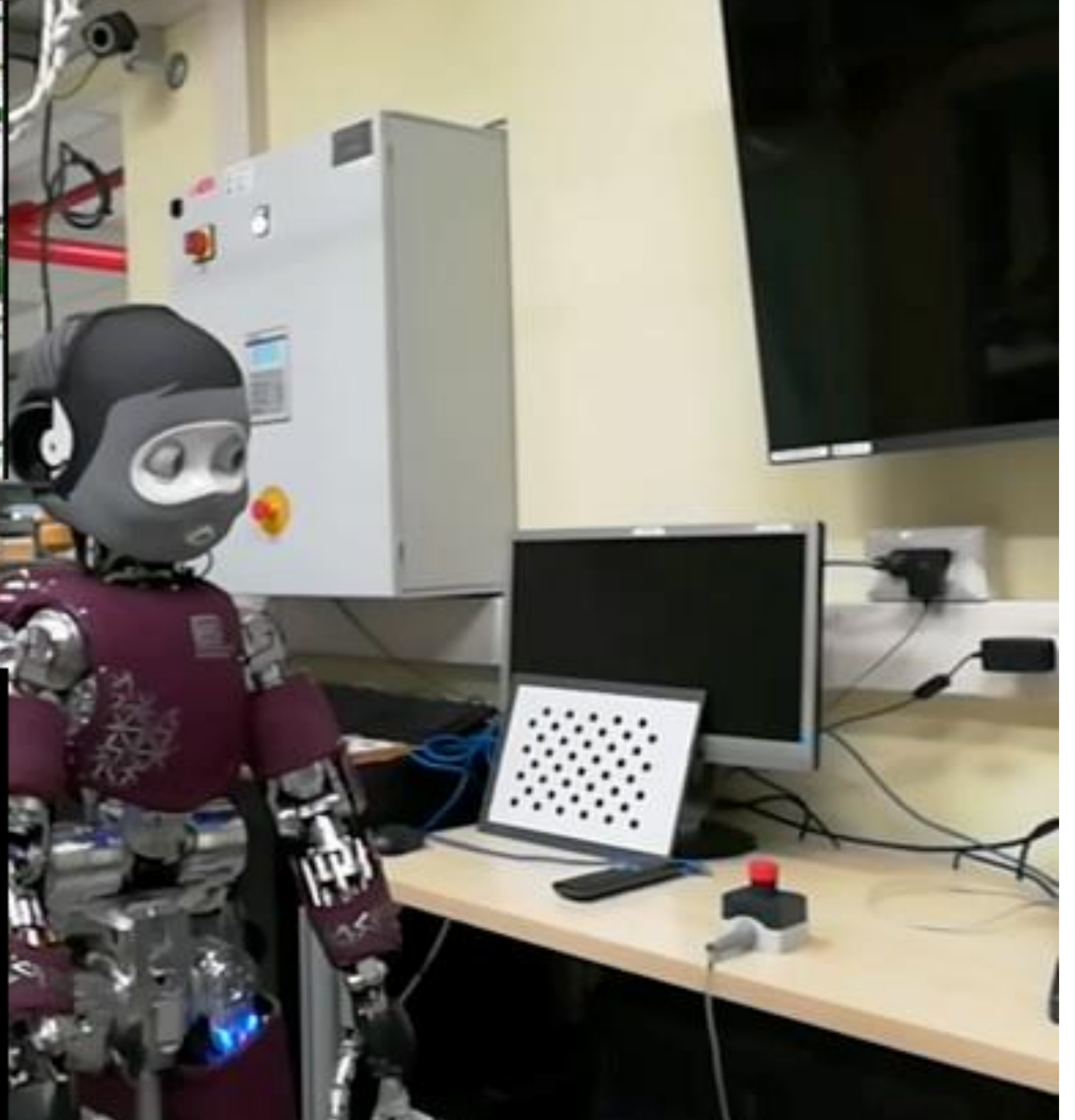
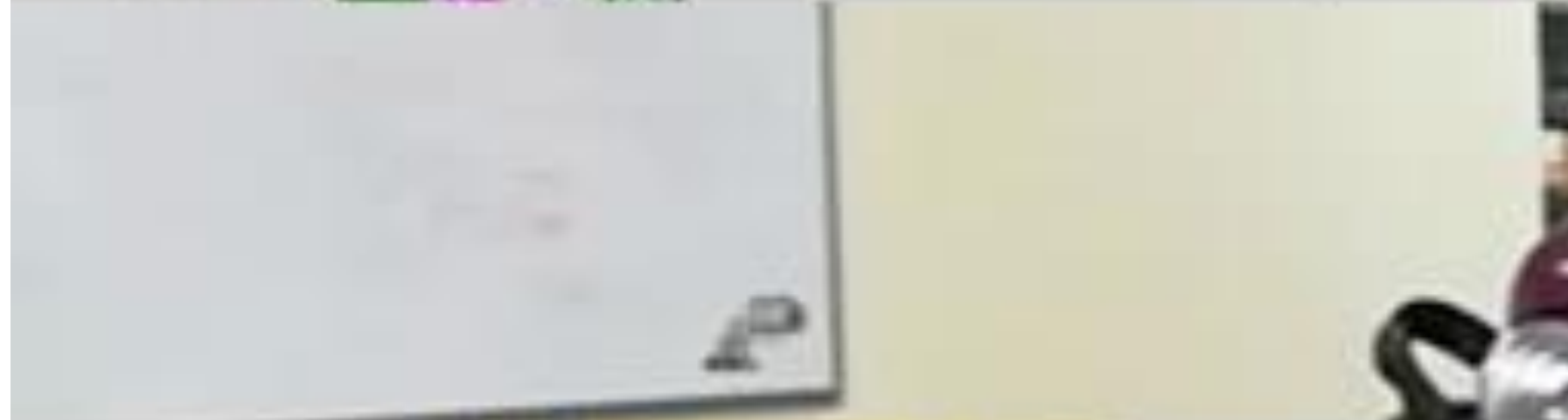
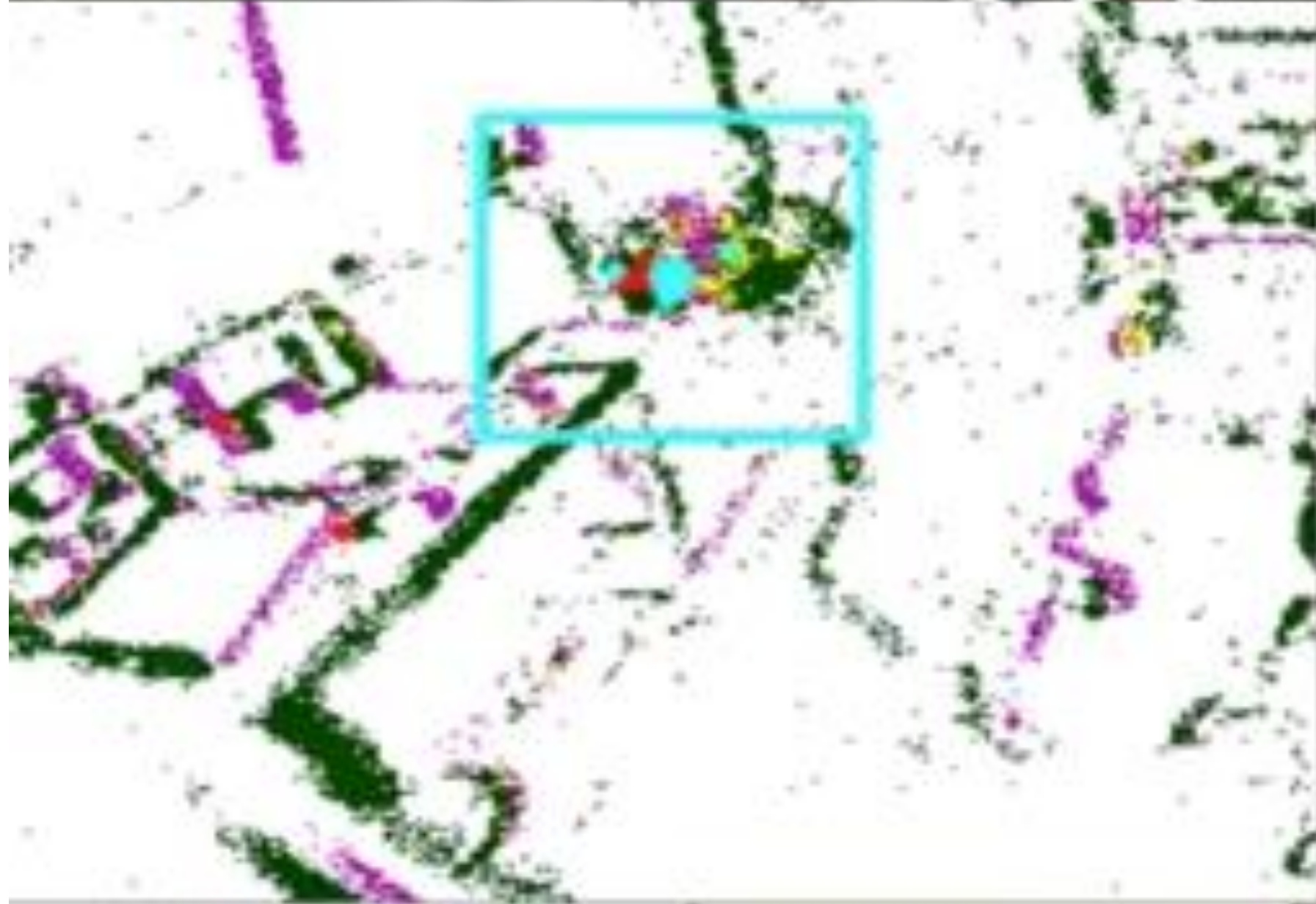
Event-driven Sensing System



Event-based Corner Detection

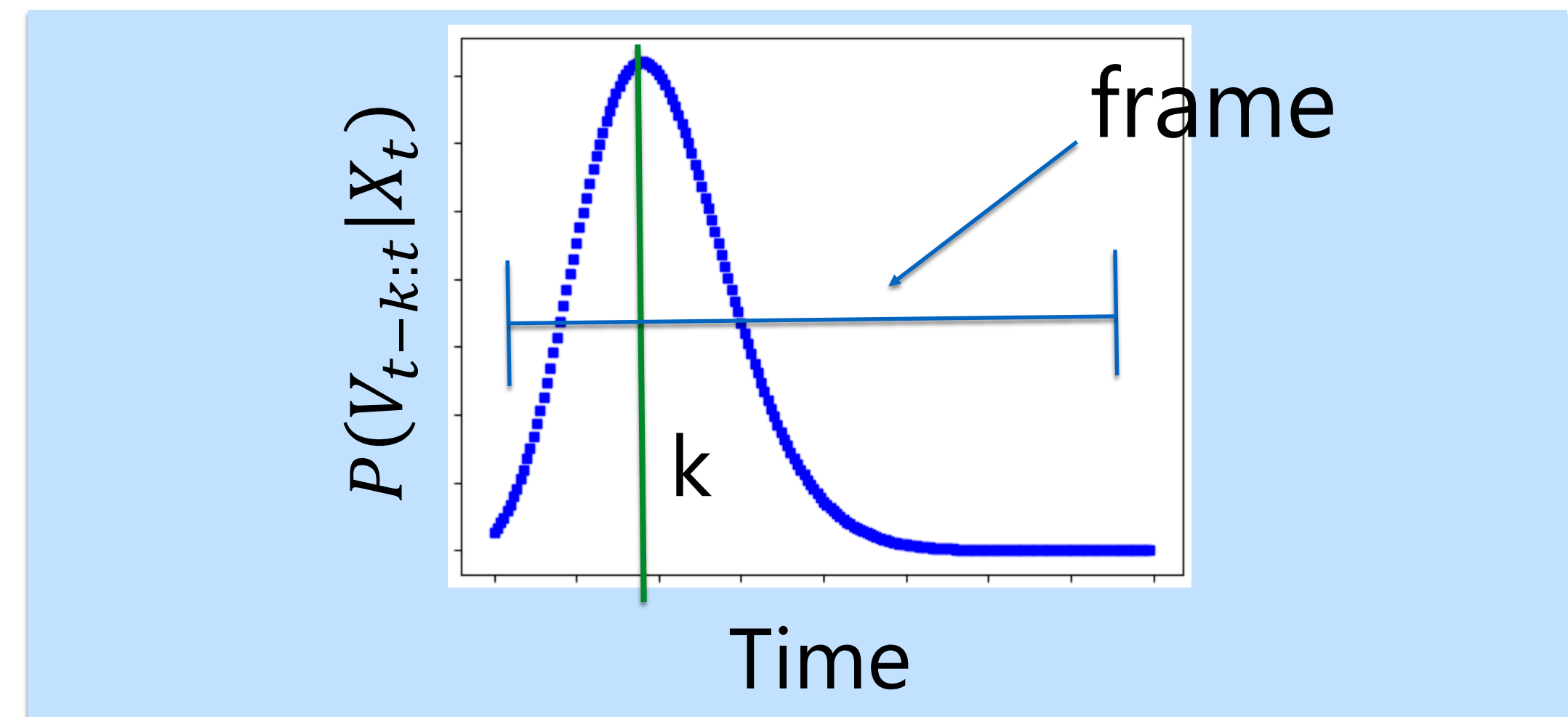
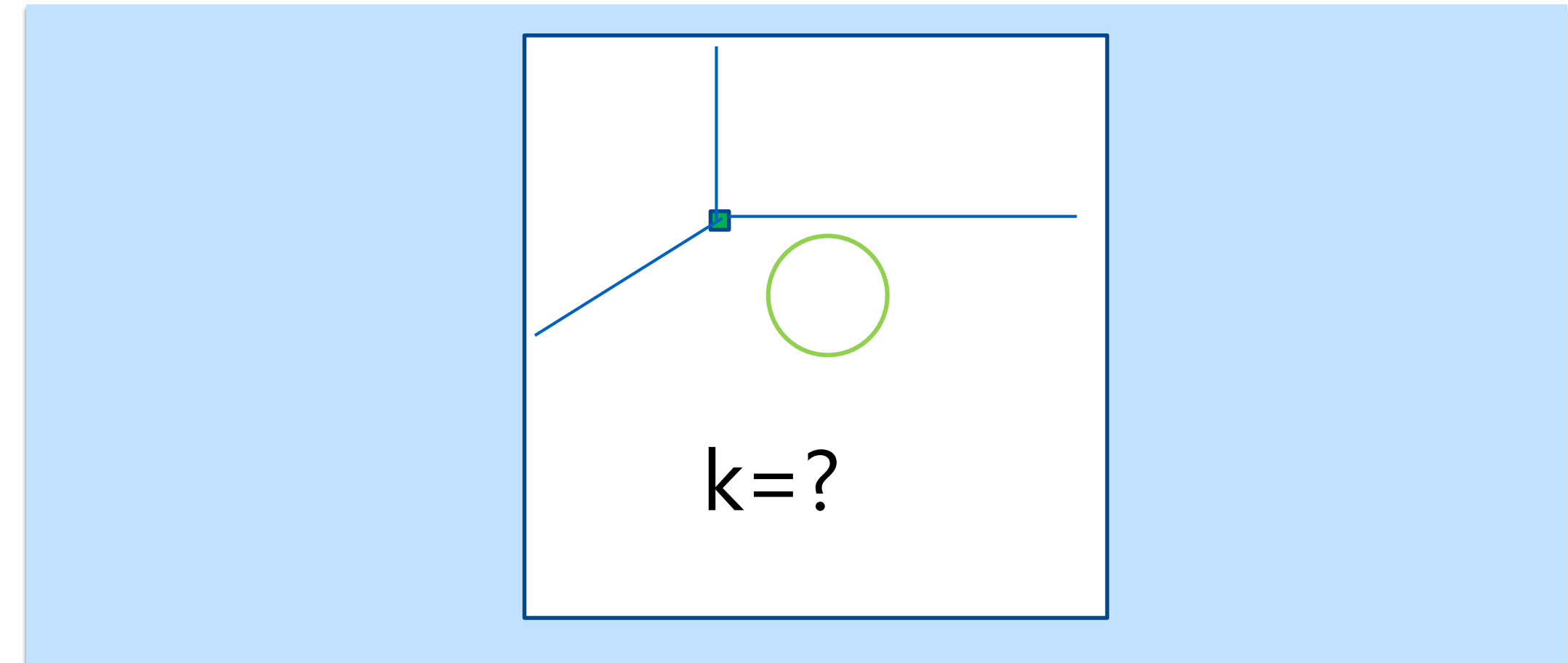
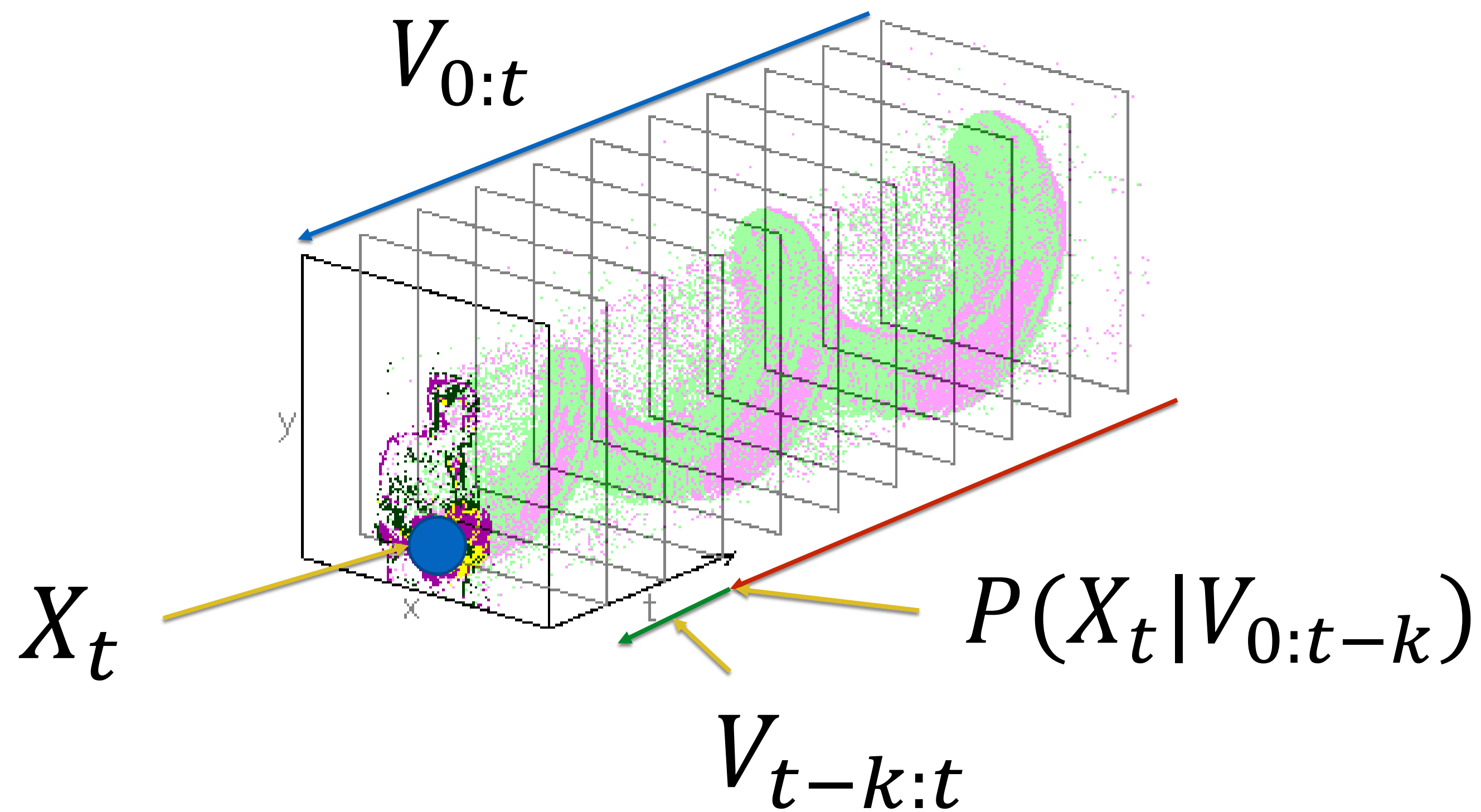
- Frame based camera:
 - frame rate: 30 Hz
 - spatial resolution: 128x128 px



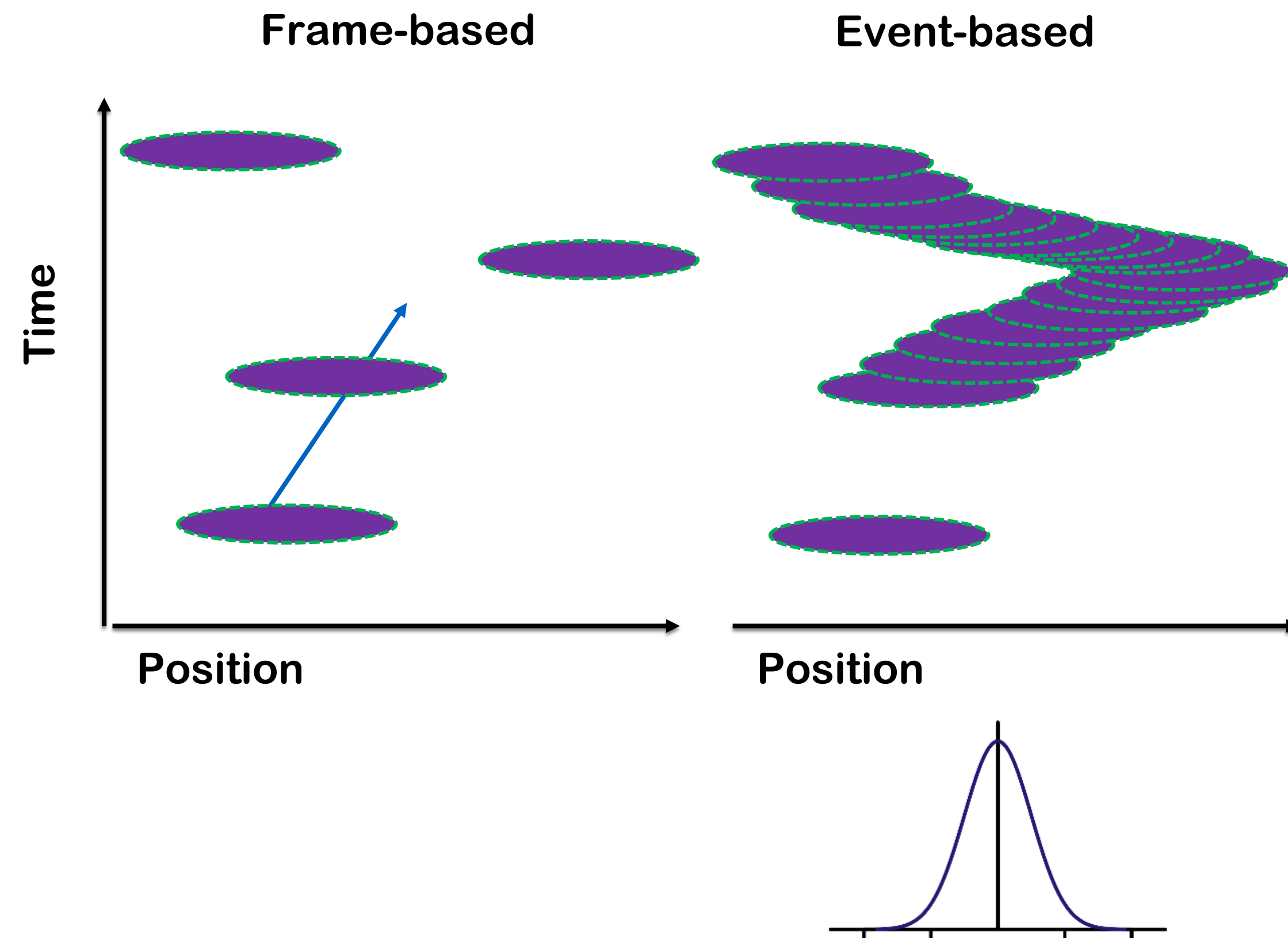


An event-driven particle filter

$$P(X_t|V_{0:t}) = \frac{P(V_{t-k:t}|X_t)P(X_t|V_{0:t-k})}{P(V_{t-k:t})}$$

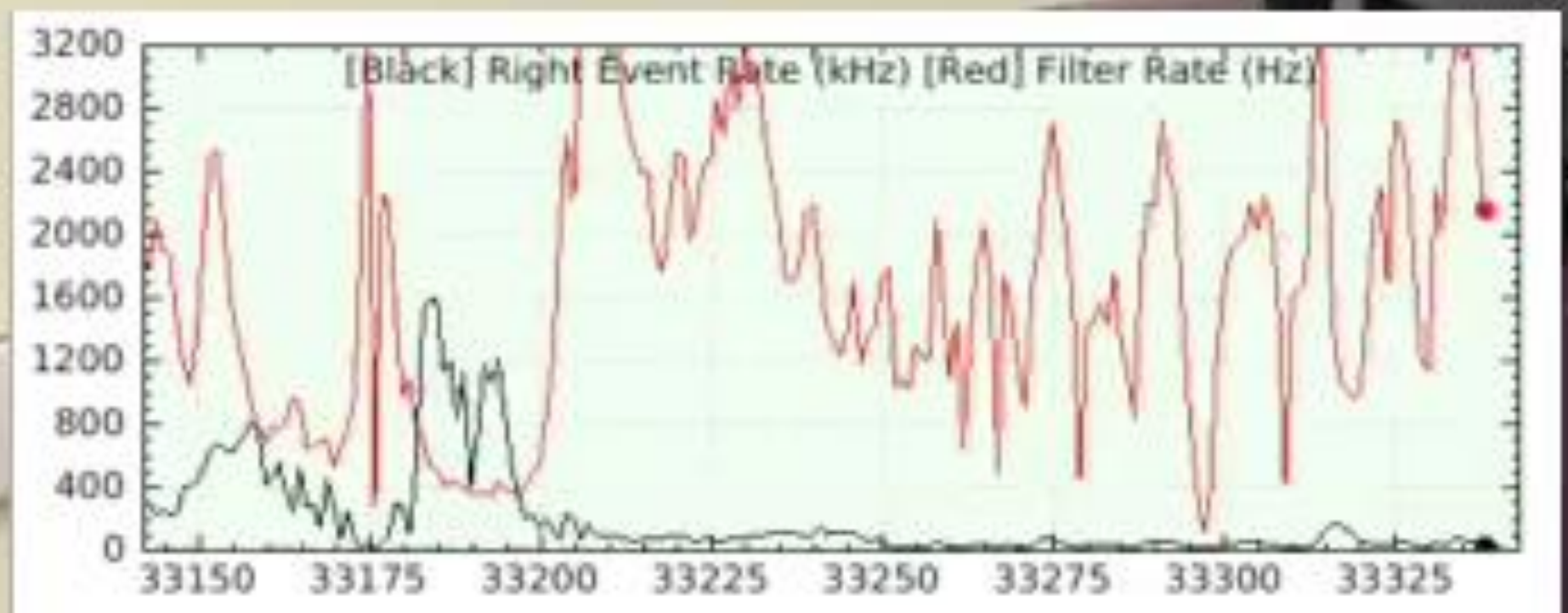
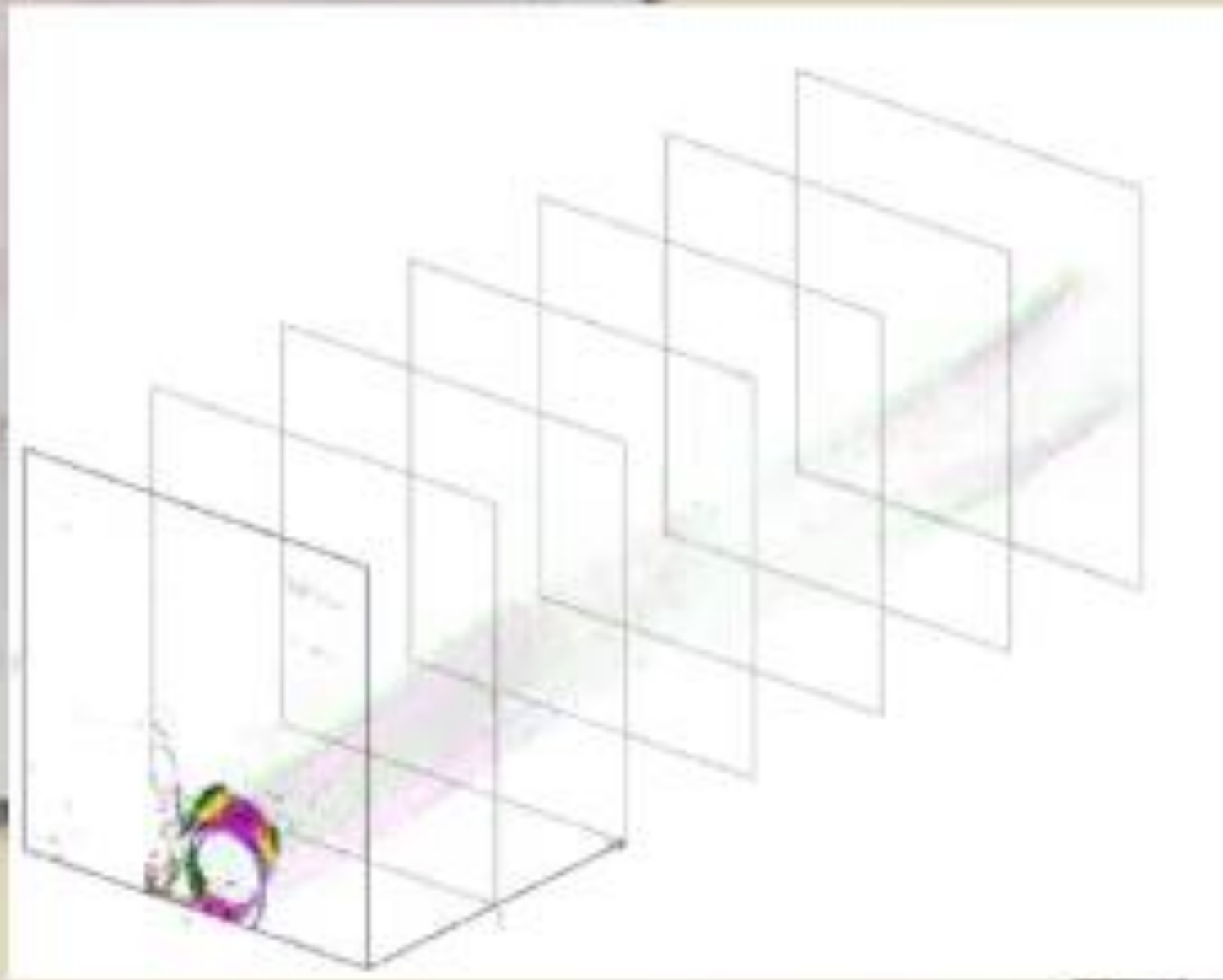


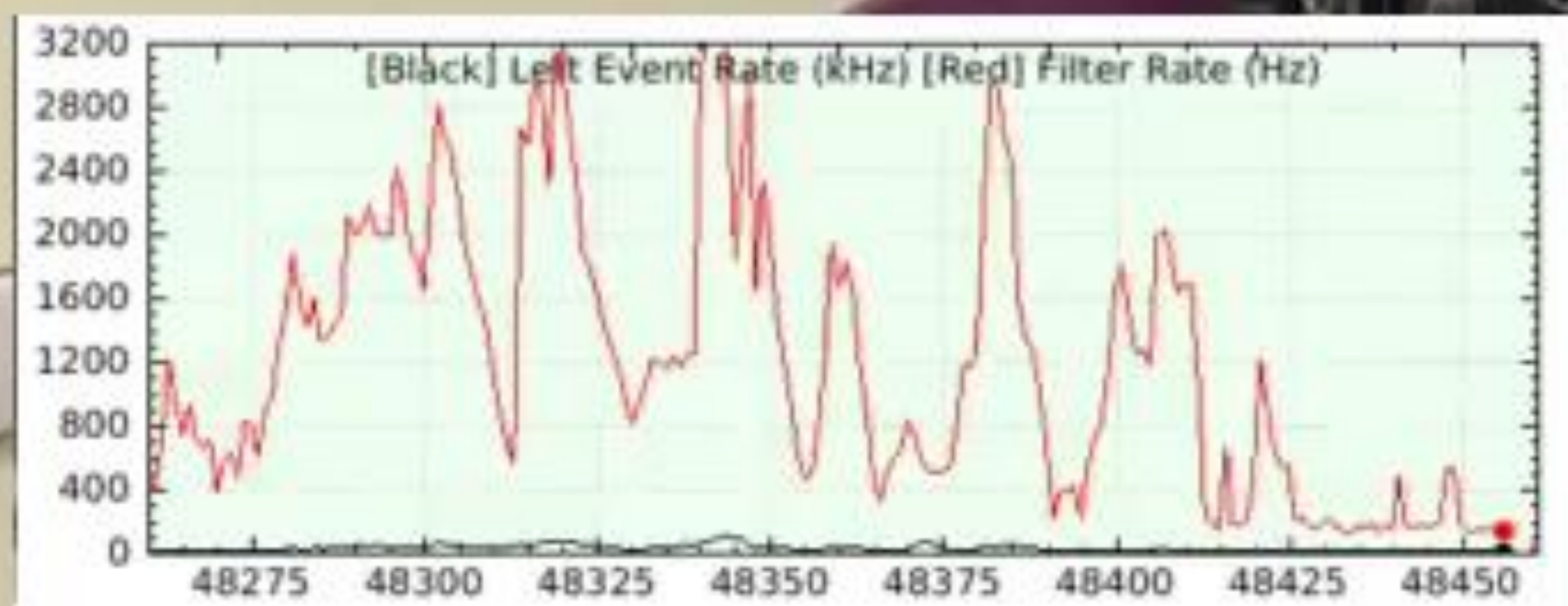
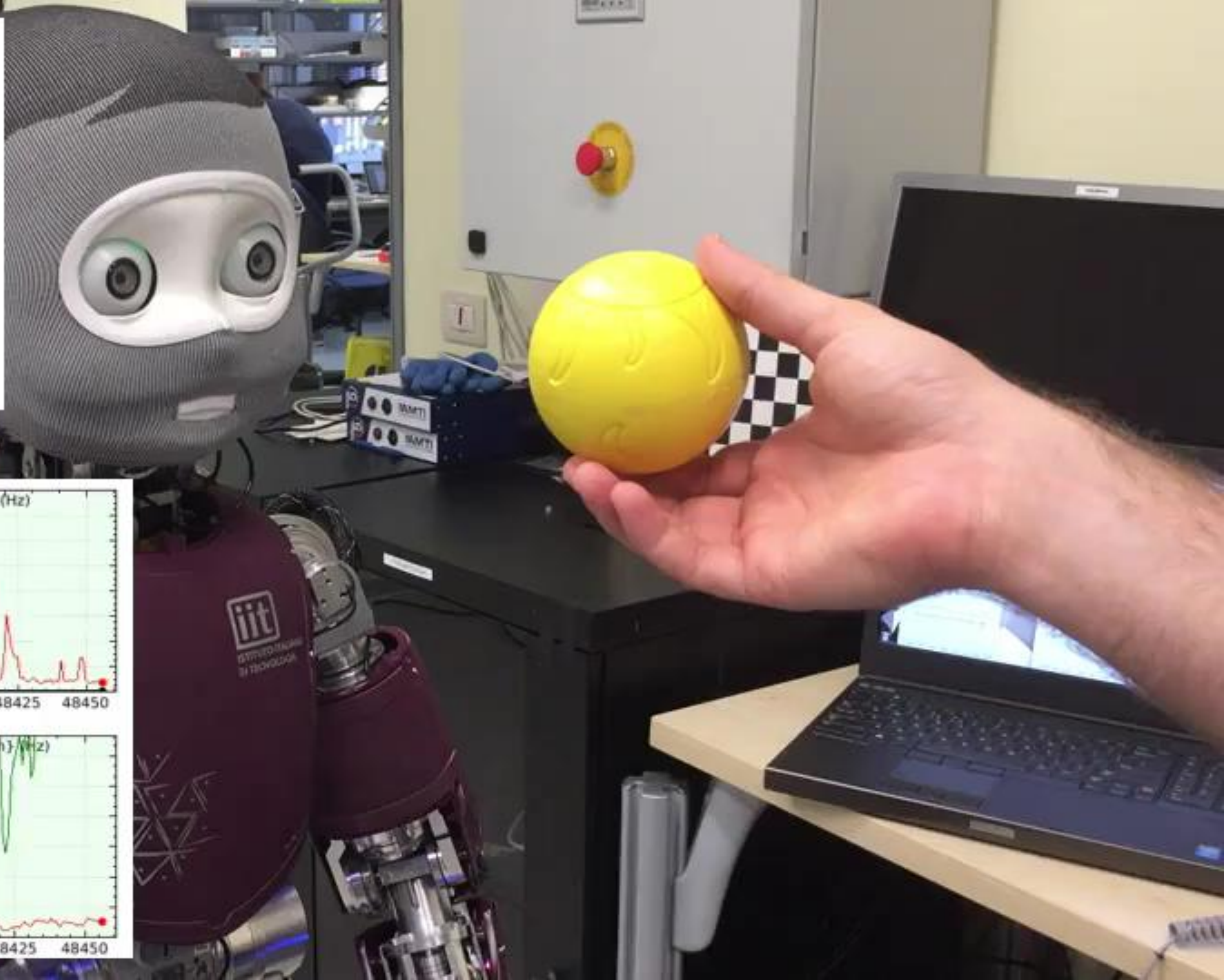
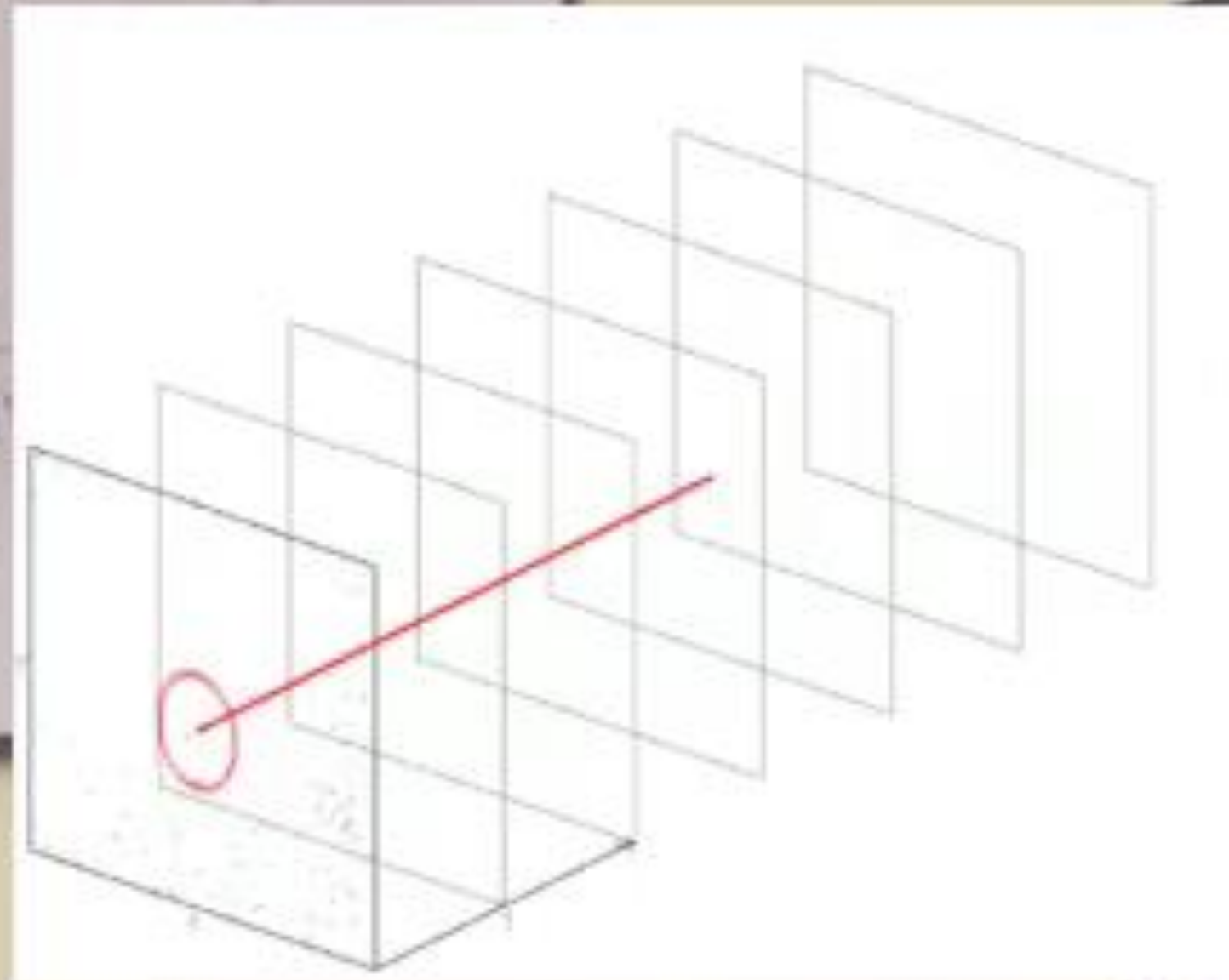
No motion model needed!



A constant position model can be used

- [1] A. Glover and C. Bartolozzi, "Robust Visual Tracking with a Freely-moving Event Camera," in *IEEE International Conference on Intelligent Robots and Systems*, 2017, pp. 3769–3776.

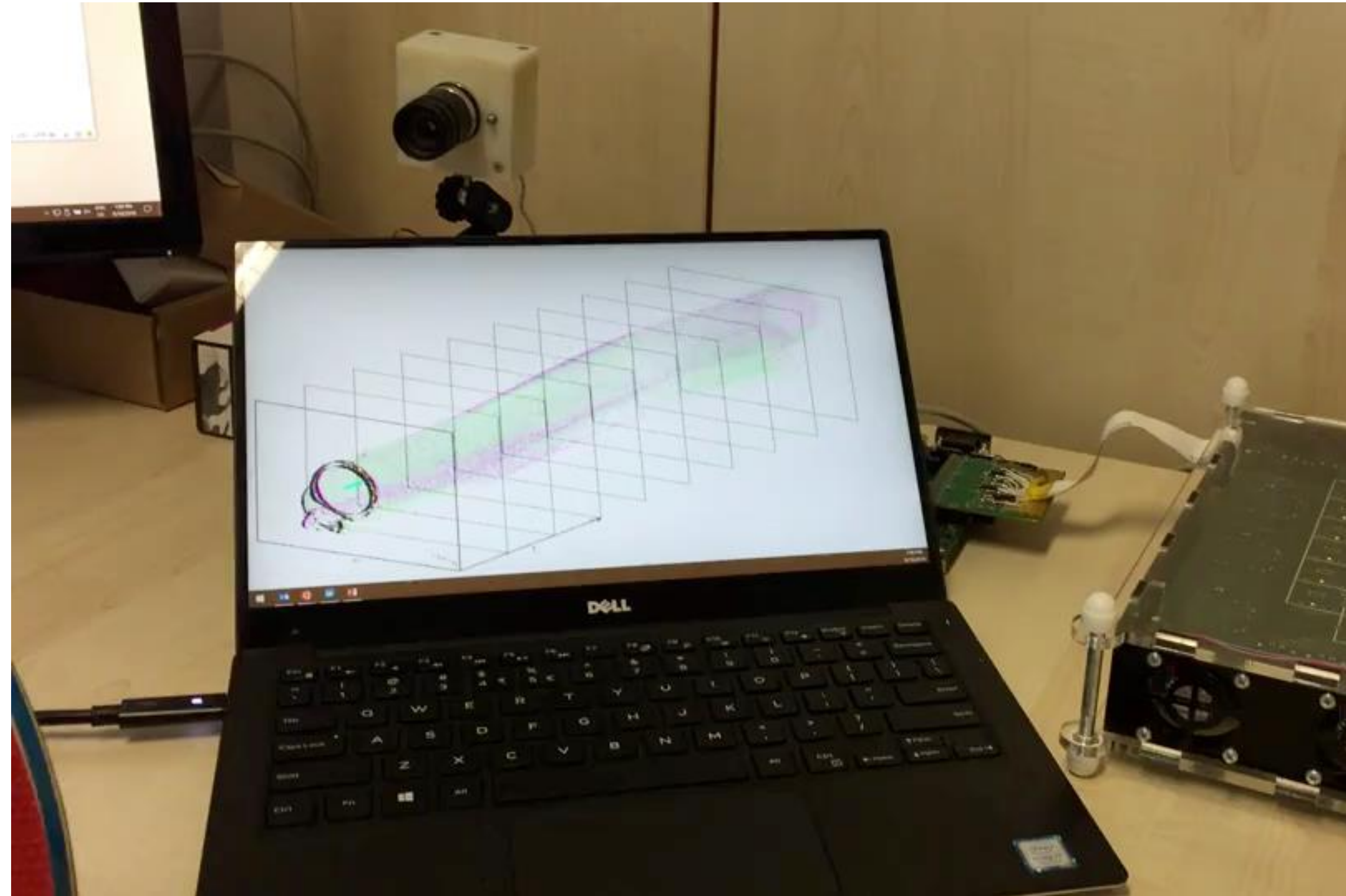
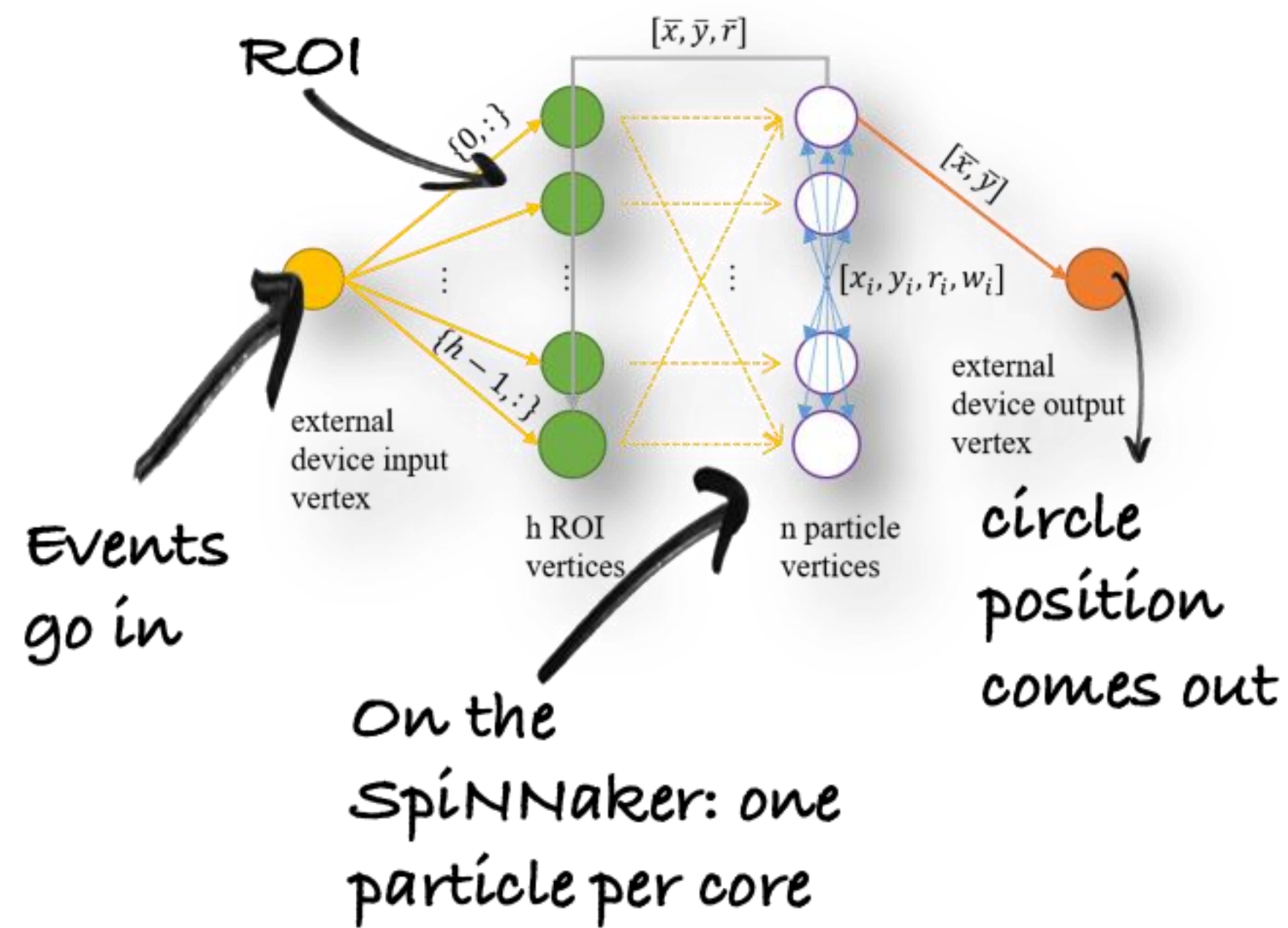




Particle Filter on SpiNNaker



- 865 ARM cores
- Asynchronous
- Parallel



Problems to be solved for a pervasively neural processing

- Spiking models:
 - Stereo disparity with a SNN
 - Head pose estimation with Head-direction networks
- Neural algorithms
 - Scalable and flexible
 - Accessible hardware
 - Integration with common robot middle-ware (ROS/YARP)

IIT - Event-driven Perception for Robotics

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