tinyML. Talks

Enabling Ultra-low Power Machine Learning at the Edge

"Mudra: Reinventing Tiny Interfaces for Ubiquitous Digital Interaction"

Leeor Langer – Co-founder and CTO, Wearable Devices Oleksandr Pirotskiy – Team Leader of Firmware Development, Wearable Devices

August 15, 2023



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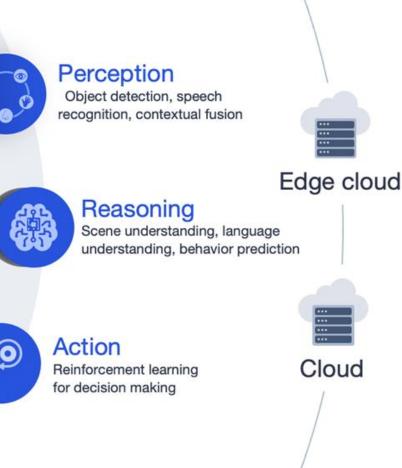
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Where what if becomes what is.

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The Leading Development Platform for Edge ML

edgeimpulse.com



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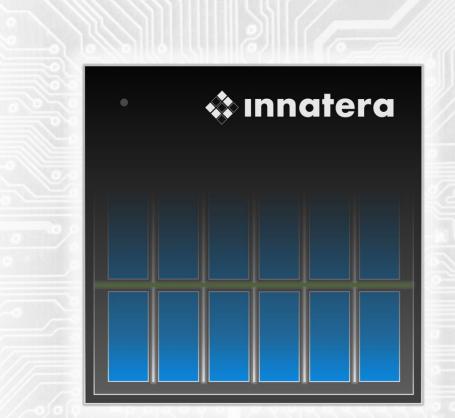
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tinyML - Enabling ultra-low Power ML at the Edge

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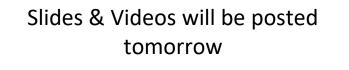
2023 Edge Al Technology Report

The guide to understanding the state of the art in hardware & software in Edge Al.





Reminders







tinyml.org/forums youtube.com/tinyml



Please use the Q&A window for your questions





Leeor Langer



Leeor Langer serves as a seasoned data scientist and the co-founder and CTO of Wearable Devices. He has been leading the company's bio-potential sensing algorithms from a garage initiative all the way to a successful Initial Public Offering (IPO) on NASDAQ. Before his venture with Wearable Devices, Mr. Langer held key roles driving algorithmic developments in numerous organizations across diverse modalities, including content ranking and X-ray applications. He earned a BSc in Electrical Engineering from the Technion - Israel Institute of Technology and an MSc in Applied Mathematics from Tel Aviv University.



Oleksandr Pirotskiy



Oleksandr Pirotskiy is the Team Leader of Firmware Development at Wearable Devices Ltd, specializing in the field of embedded software development. Oleksandr is currently focused on the development of gesture recognition algorithms, bringing extensive experience in the development and integration of sensor drivers. His work is distinguished by his innovative approach to low-power solutions, ensuring the efficient use of resources without compromising functionality. Oleksandr has a proven track record in implementing a substantial number of HID devices and ensuring their seamless integration across multiple operating systems. He holds a Master's degree in Computer Sciences from the Kiev National Polytechnic Institute. His unique combination of academic rigor, passion for technological innovation, and dedication to energy efficiency continues to propel him forward in the wearable devices industry.

Mudra: Reinventing tiny Interfaces for Ubiquitous Digital Interaction



Content



Human Computer Interaction

- Examples of digital interaction
- Today\Tomorrow
- Ionic Exchange
 - Biopotentials
 - Sensing
- Sensor Fusion
 - Data Acquisition
 - Neural Networks
 - Random Forests

Software implementation

- System overview
- Interaction within Apple Ecosystem

HCI – Today & Tomorrow



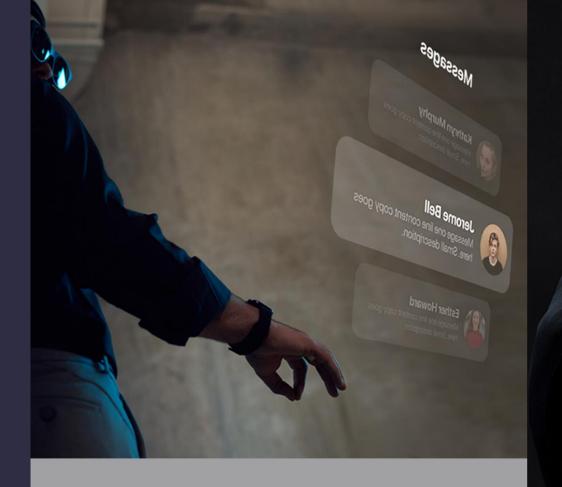


HCI – Today & Tomorrow





HCI – Today & Tomorrow



No limitation

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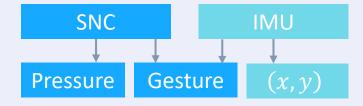


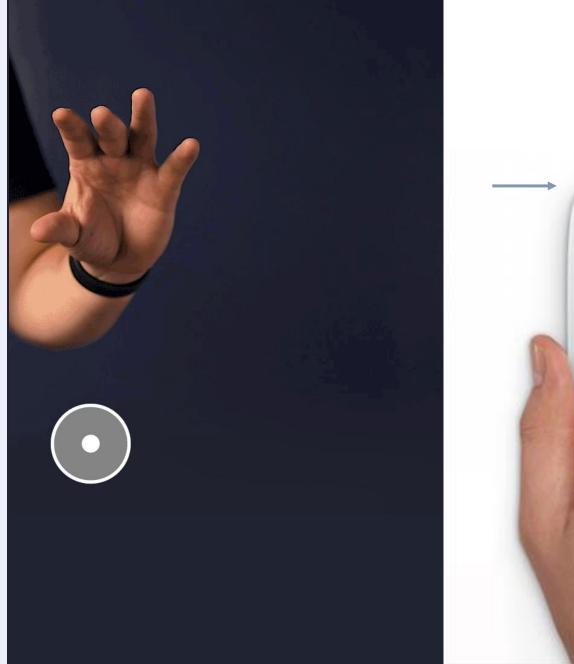
Mudra Band -Neural Clicker-Fingertip Pressure





Neural Clicker



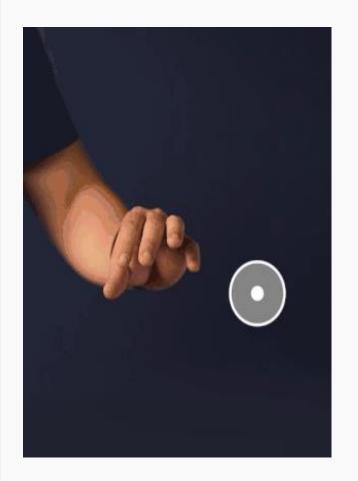


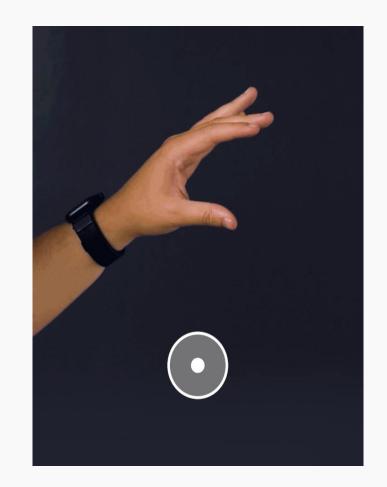




Drag & Drop -SNC Fingertip Pressure Estimation









Biopotentials

01.

What is a bio-potential signal?

Biopotentials are electrical signals generated by the body's cells, known as excitable cells. On our arm, such cells can be found in the nervous and muscular systems.

02.

What is a biopotential sensor?

A mechanism designed to detect and respond to extremely low-voltage electrical fields produced by biological processes within the human body.

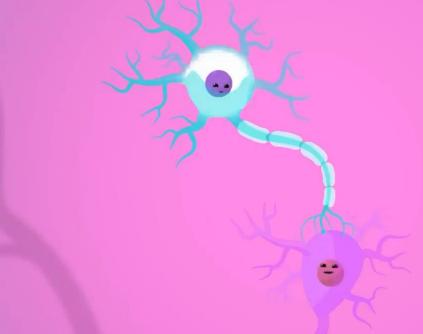


Biopotentials – Ionic Exchange

Direct glimpse into the mechanics of our body

* Anatomy and Physiology



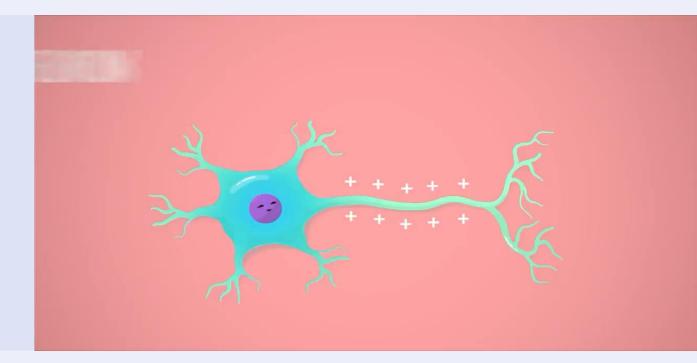




Three stages:

- 1. Action Potential Formation
- 2. Propagation
- 3. Innervation

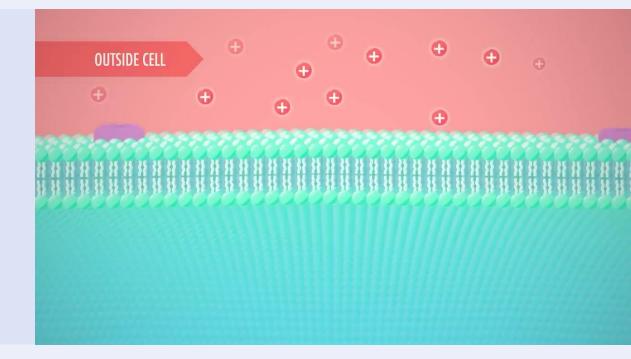
See Anatomy and Physiology





Biopotentials AP Initiation (1a)

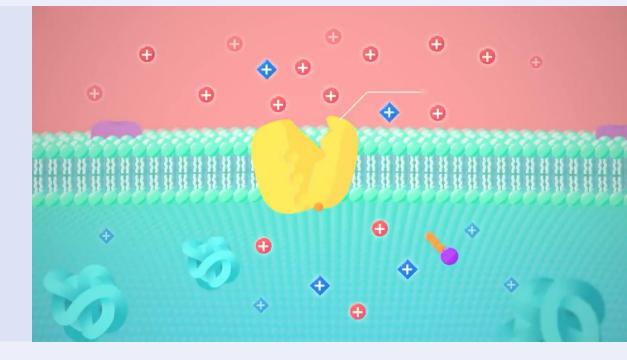
- a. Resting potential
- b. Rising phase (depolarization)
- c. Falling phase (repolarization)
- d. Hyperpolarization





Biopotentials AP Initiation (1a)

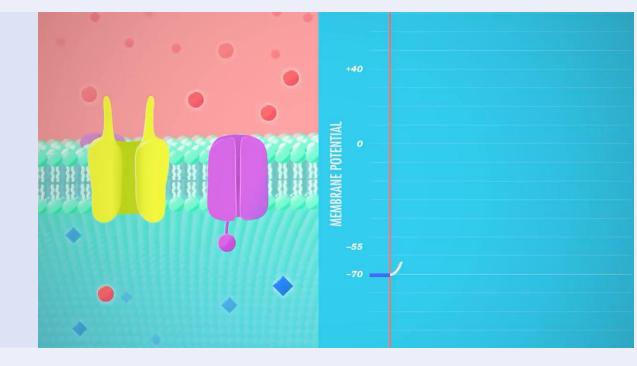
- a. Resting potential
- b. Rising phase (depolarization)
- c. Falling phase (repolarization)
- d. Hyperpolarization





Biopotentials AP Initiation (1b)

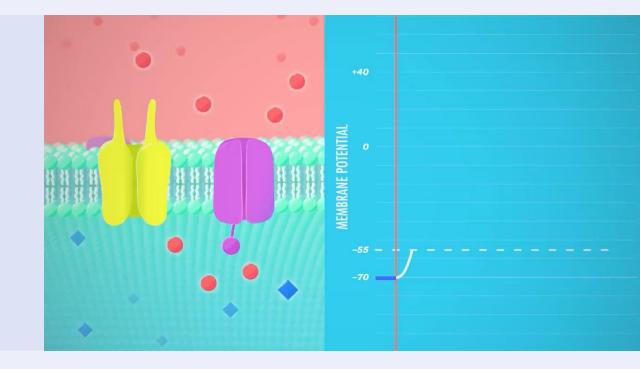
- What is an Action Potential?
- Resting potential
- Rising phase (depolarization)
- Falling phase (repolarization)
- Hyperpolarization





Biopotentials AP Initiation (1b)

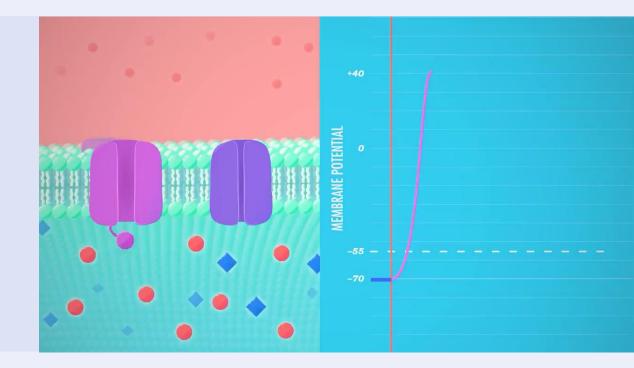
- Resting potential
- Rising phase (depolarization)
- Falling phase (repolarization)
- Hyperpolarization





Biopotentials AP Initiation (1c)

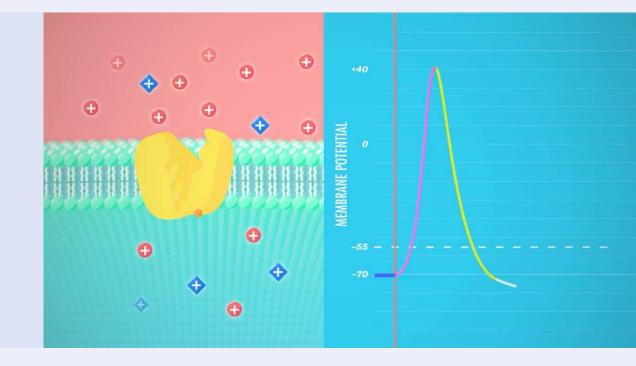
- Resting potential
- Rising phase (depolarization)
- Falling phase (repolarization)
- Hyperpolarization





Biopotentials AP Initiation (1d)

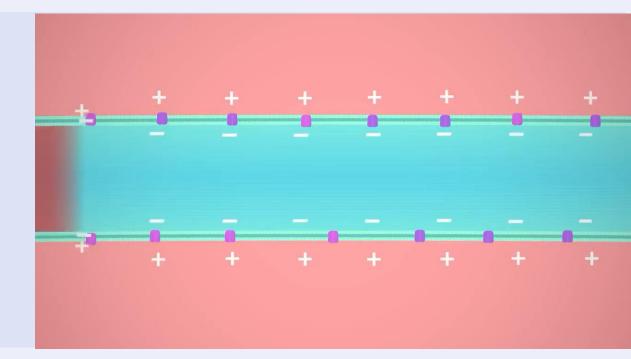
- Resting potential
- Rising phase (depolarization)
- Falling phase (repolarization)
- Hyperpolarization





Biopotentials Propagation (2)

The Action Potential propagates like a wave!





Biopotentials Innervation (3)

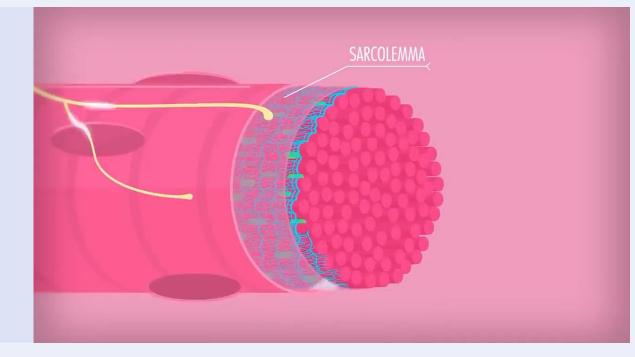
 $\mathsf{Electrical} \rightarrow \mathsf{Chemical}$





Biopotentials Muscle Contraction (3)

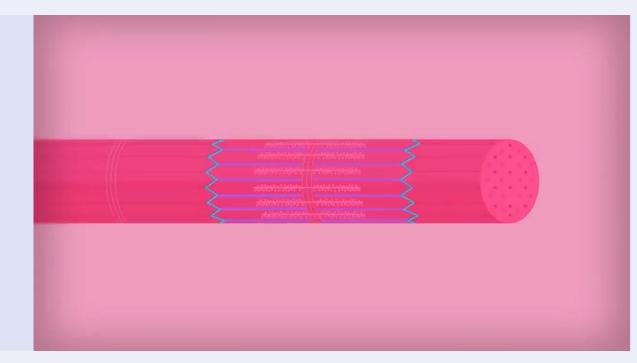
Another conversion from electrical \rightarrow chemical \rightarrow mechanical





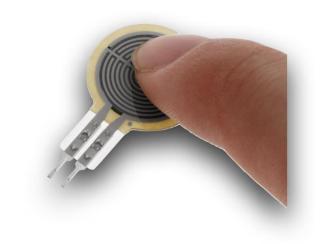
Biopotentials Muscle Contraction (3)

Our body is like a *marionette*, our muscles contract and pull on the tendons connected to the bone structure.





Data Acquisition







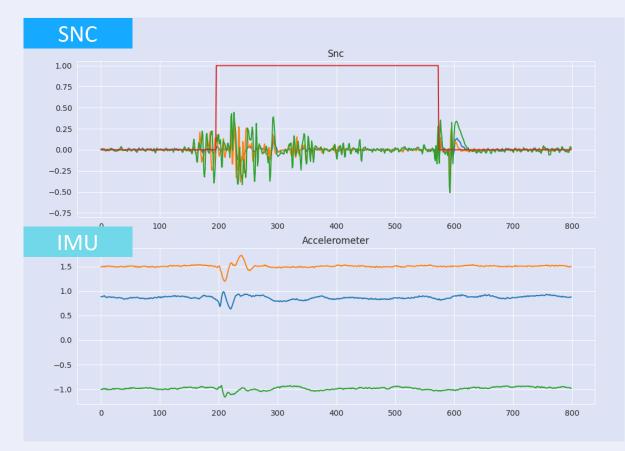


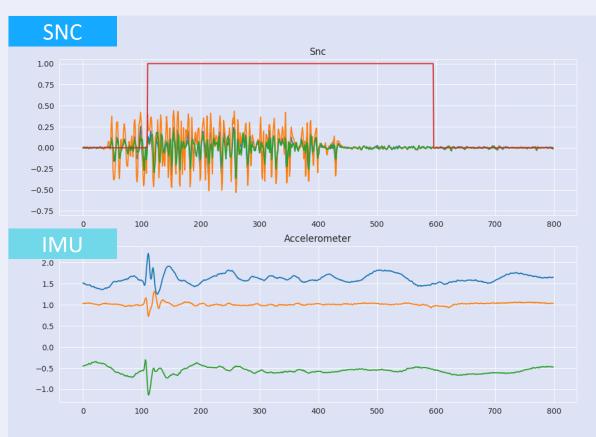
- In order to collect data we need an API into our nervous system?
- Manual segmentation was our first method!

• Technion IIT

• Automated labeling with a conductive fabric based "button"

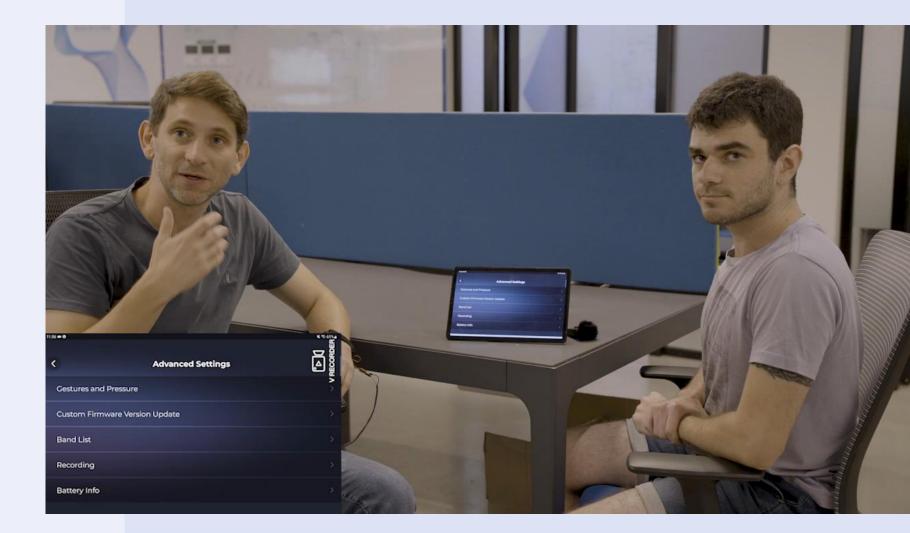
Examples: Tap







Data Acquisition Real-time

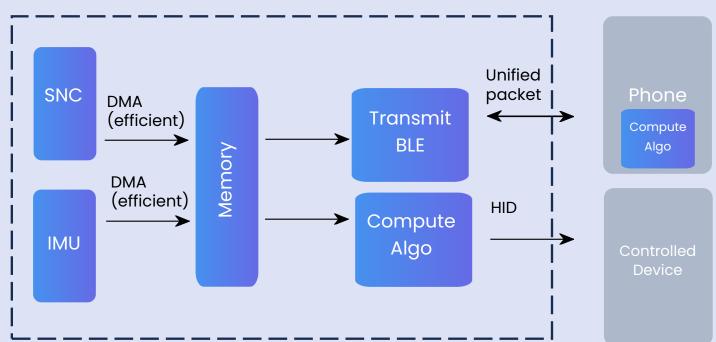




System Optimization

- Queuing mechanism with additional allocation to prevent data corruption.
- Decision tree to make dynamic prioritize, what to send.
- Adaptation of gap and gatts parameters : connection interval, event length , data length, MTU, etc.
- Reducing cpu usage.

CPU (FW)





01.

What is a bio-potential signal?

Biopotentials are electrical signals generated by the body's cells, known as excitable cells. On our arm, such cells can be found in the nervous and muscular systems.

02.

What is a biopotential sensor?

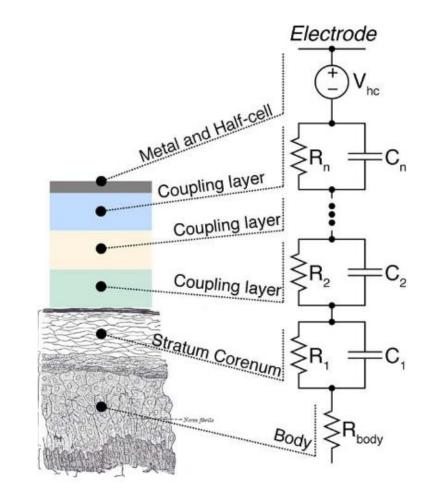
A mechanism designed to detect and respond to extremely low-voltage electrical fields produced by biological processes within the human body.



02.

What is a biopotential sensor?

A mechanism designed to detect and respond to extremely low-voltage electrical fields produced by biological processes within the human body.



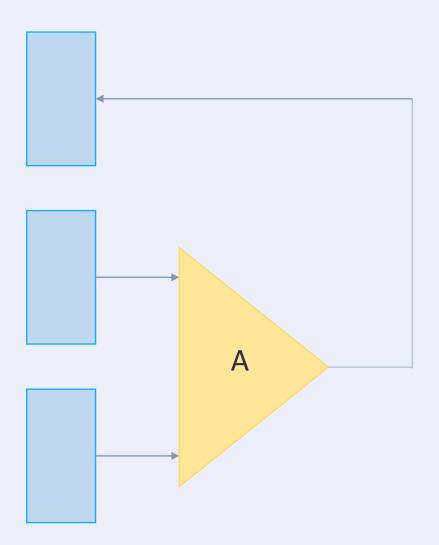
* Anatomy and Physiology <u>link</u>



02.

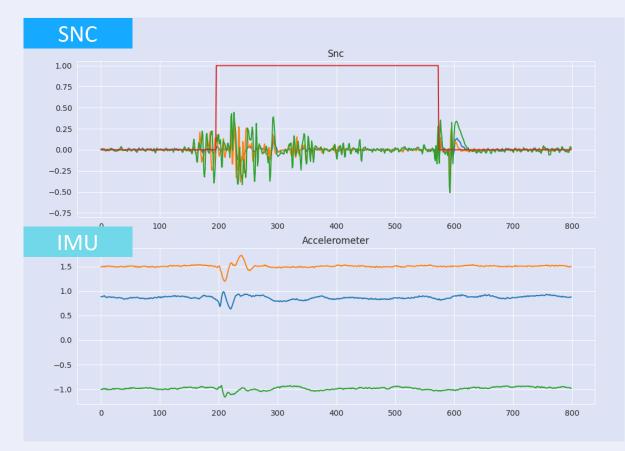
What is a biopotential sensor?

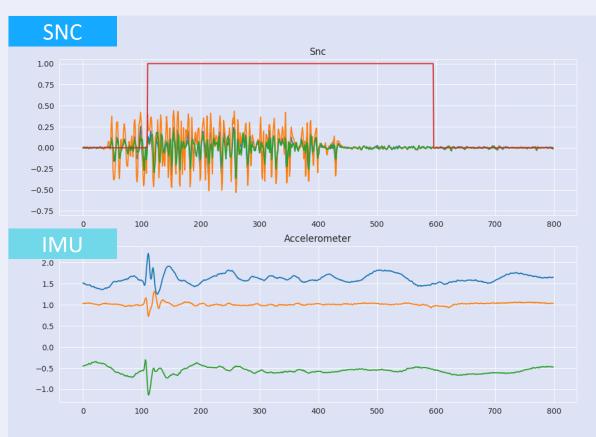
A mechanism designed to detect and respond to extremely low-voltage electrical fields produced by biological processes within the human body.





Examples: Tap







Examples: Noise







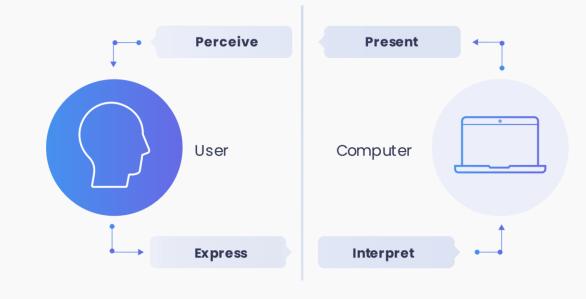
Biopotentials vs Motion Sensors

Wearable
Devices

	IMU	SNC
Latency	Х	\checkmark
Power	Х	\checkmark
Fingertip-Vibration	\checkmark	\checkmark
Fingertip-Pressure	Х	\checkmark
Skin Contact	\checkmark	Х

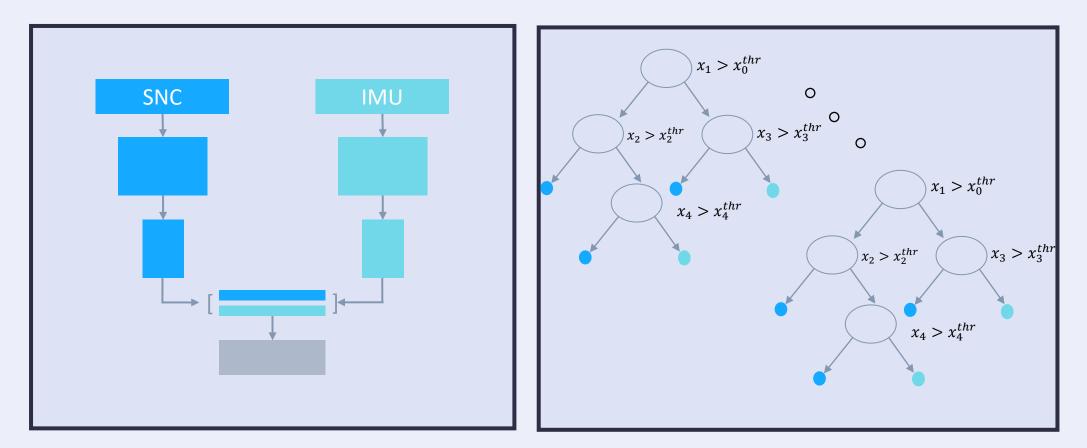
HCI – Keyboard Press Review





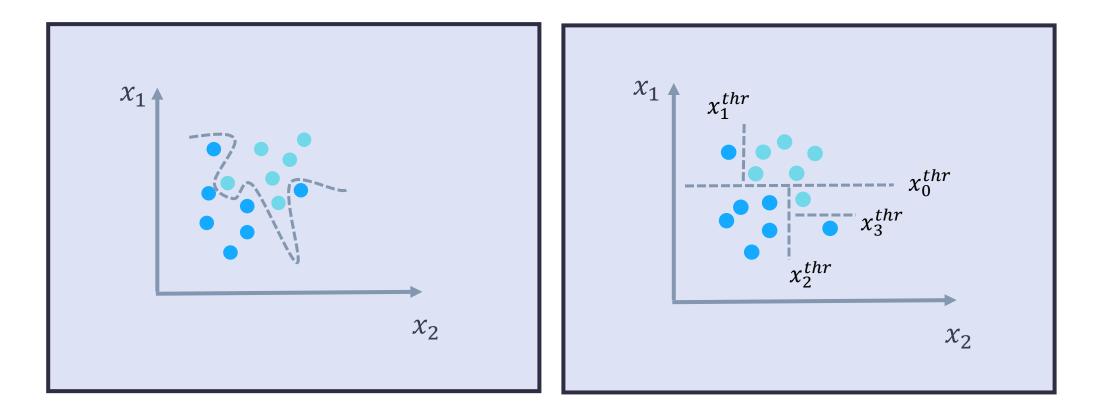


Sensor Fusion NN \ RF





Sensor Fusion NN \ RF





Gesture Recognition

Variability between users:

- Usage characteristics
- Physiological characteristics
- Device characteristics



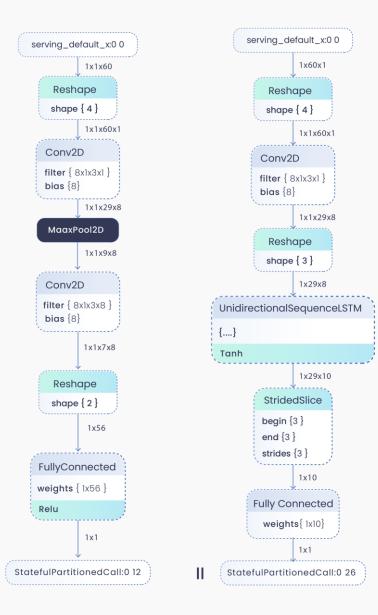
Variability between different gestures and noise:

- Unique noise sources
- No off-the-shelf data
- Computational constraints



Gesture Recognition MobileNets vs tinyNN

Network	Тор 1	Params	MAdd	CPU
MobileN et VI	70.6	4.2M	575M	113ms
ShuffleNet (1.5)	71.5	3.4M	292M	
ShuffleNet (x2)	73.7	5.4M	524M	
NasNet-A	74.0	5.3M	564M	183ms
MobileN et V2	72.0	3.4M	300M	75ms
MobileNetV2 (1.4)	74.7	6.9M	585M	143ms





(x, y) cursor projection

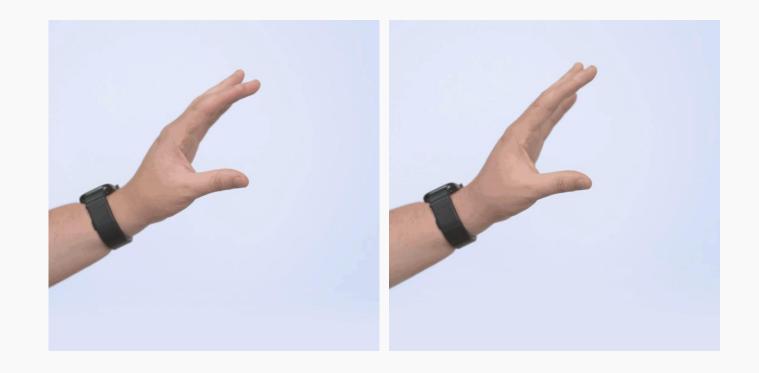


- If user wear the band not like designed ?
- What happens when we tilt our hand?
- "In the wild" we can simply reorient:

$$x = \tan^{-1} \left(acc_y, \sqrt{acc_x^2 + acc_y^2} \right),$$
$$y = \tan^{-1} \left(-acc_x, \sqrt{acc_y^2 + acc_z^2} \right)$$



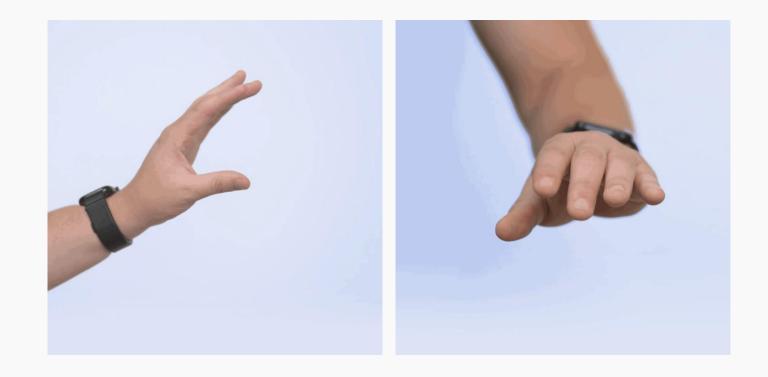
Discrete vs Continuous Movements



• Momentary tap vs drag and drop



Reactive vs Proactive

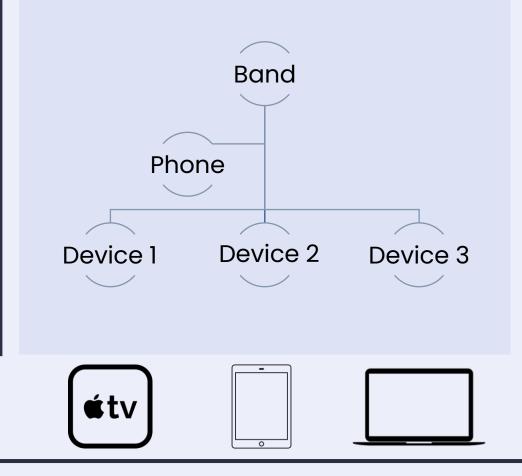


- Tap on interface HID (custom\mouse\touchpad)
- Answer a call <u>ANCS AMS</u>



2nd Device

Ble broadcast, to achieve all possible usage of gestures with all devices, without any additional software. System req direct adv, custom blacklist adv, pairing mechanism





HCI – Computer Vision vs SNC \IMU

Wearable Devices

	Computer Vision	Wearable Devices
Sensor Power Consumption	±500mW	±10mW
Compute Power Consumption	±1000mW	±10mW
Training Complexity	\checkmark	\checkmark
HUD weight	Х	\checkmark
Skin Contact	\checkmark	Х

HCI – Computer Vision vs SNC \IMU

Advantages:

- High accuracy
- Signal denoising not required
- Easy to train

Disadvantages:

- Pseudo real-time, on embedded low power device, no capability to work without latency
- Power consumption Typically consumes around ±3000ma
- Amount of training data to operate is huge

Please send any questions to the following emails: leeor.langer@wearabledevices.co.il alex.p@wearabledevices.co.il







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