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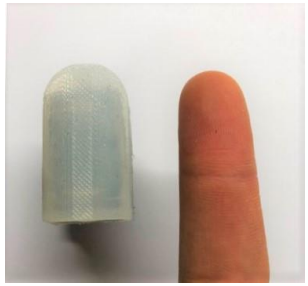
ROBOTICS SCIENCE AND SYSTEMS

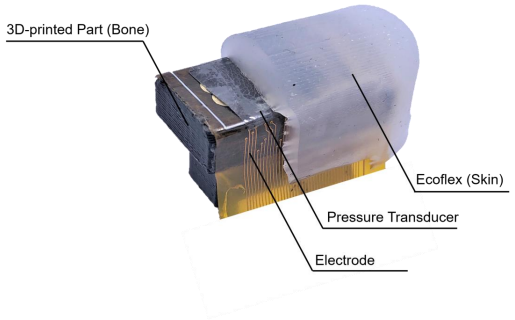
Event-Driven Visual-Tactile Sensing and Learning for Robots

Tasbolat Taunyazov, Weicong Sng, Hian Hian See, Brian Lim, Jethro Kuan, Abdul Fatir Ansari, Benjamin C.K. Tee, Harold Soh

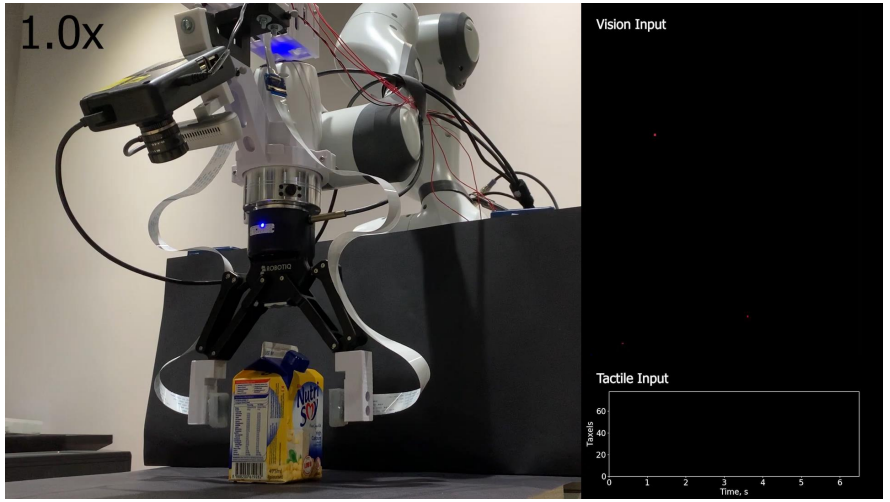
Event-Driven Visual-Tactile Sensing and Learning for Robots

1





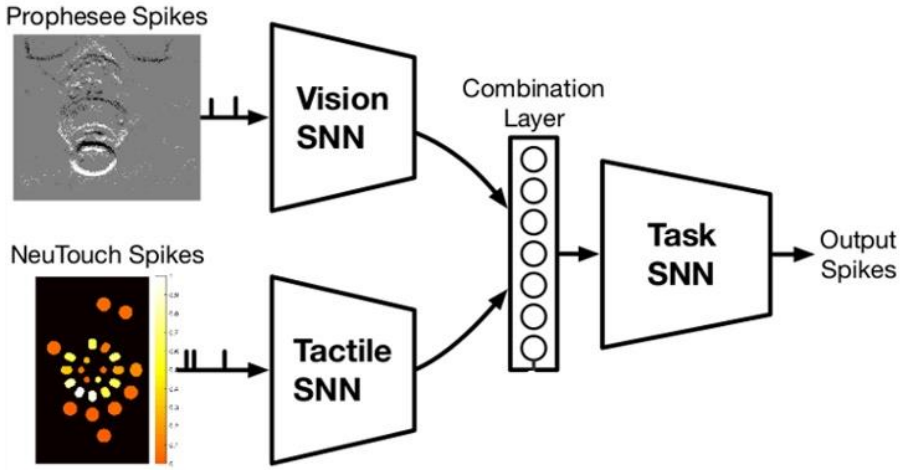
Event-based tactile sensor with graphene-based piezoresistive transducer



VT-SNN is tested for two different robotic tasks:

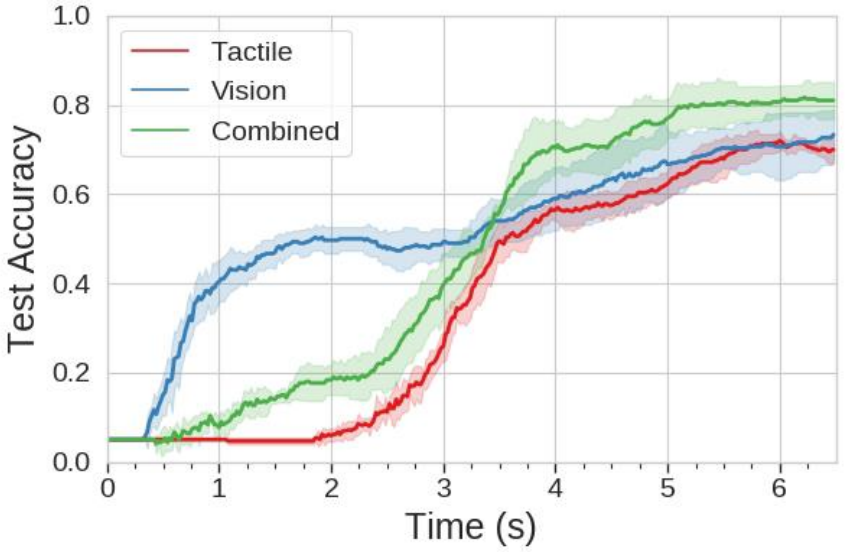
- Container and weight classification
- Rotational slip detection

2



Visual-Tactile Spiking Neural Network

$$L_\omega = \frac{1}{2}\sum_{n=0}^{N_o}\left(\sum_{t=0}^{T}\omega(t)s^n(t) - \sum_{t=0}^{T}\omega(t)\tilde{s}^n(t)\right)^2$$



We have achieved high early classification accuracy with proposed loss.

Multimodal datasets are publicly available at https://clear-nus.github.io/visuotactile/

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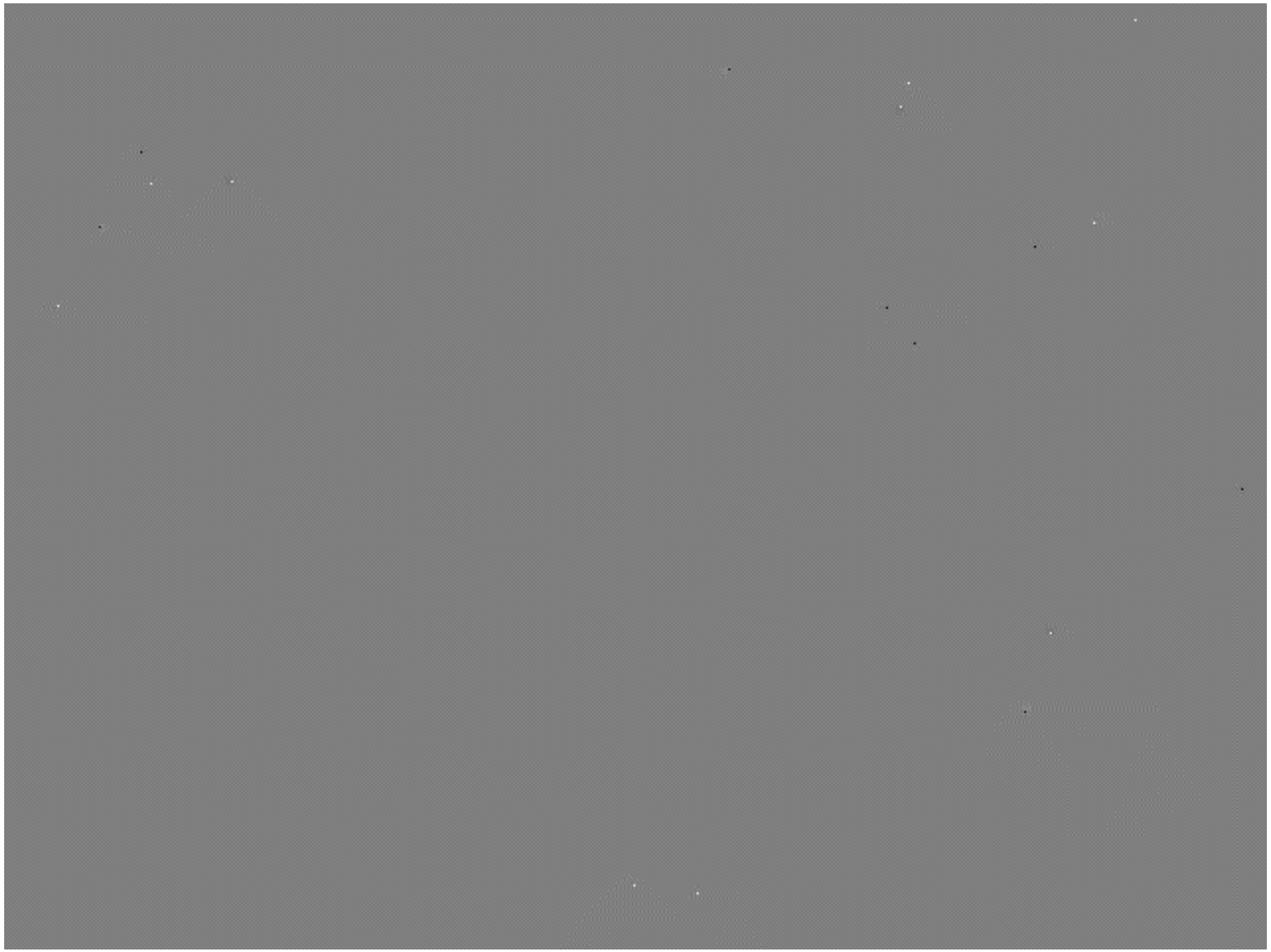
Event-Based Camera



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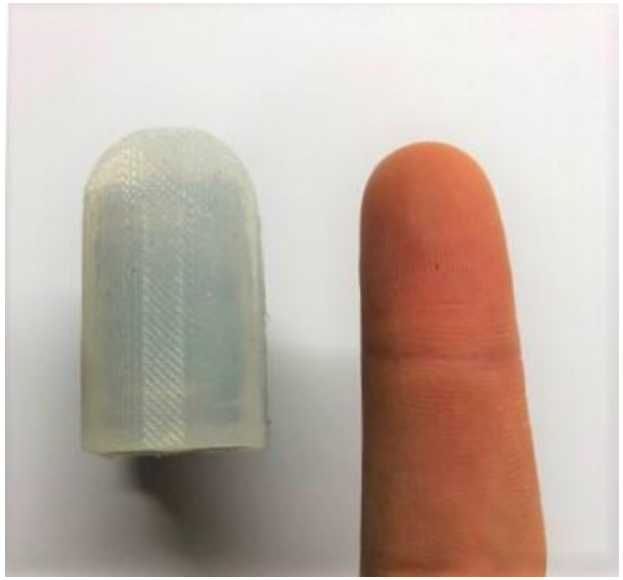
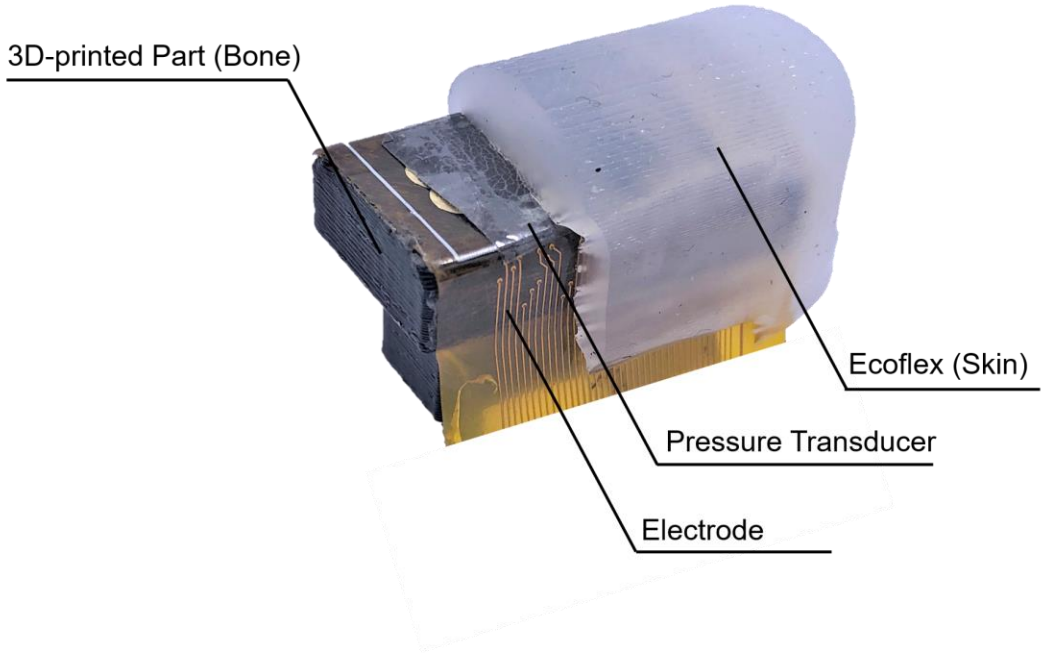
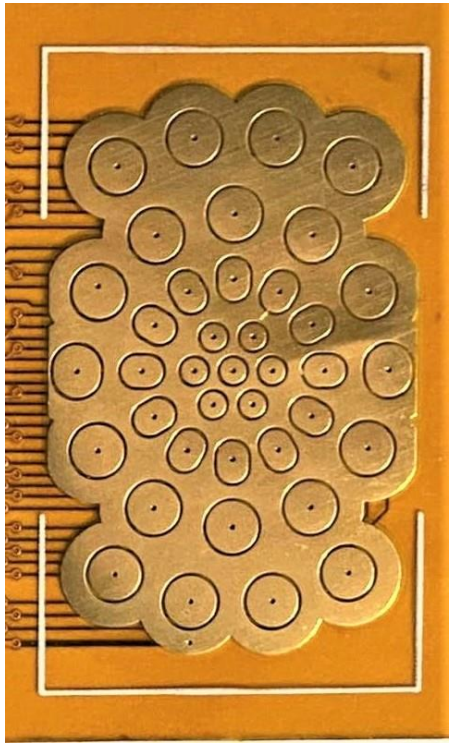
NeuTouch







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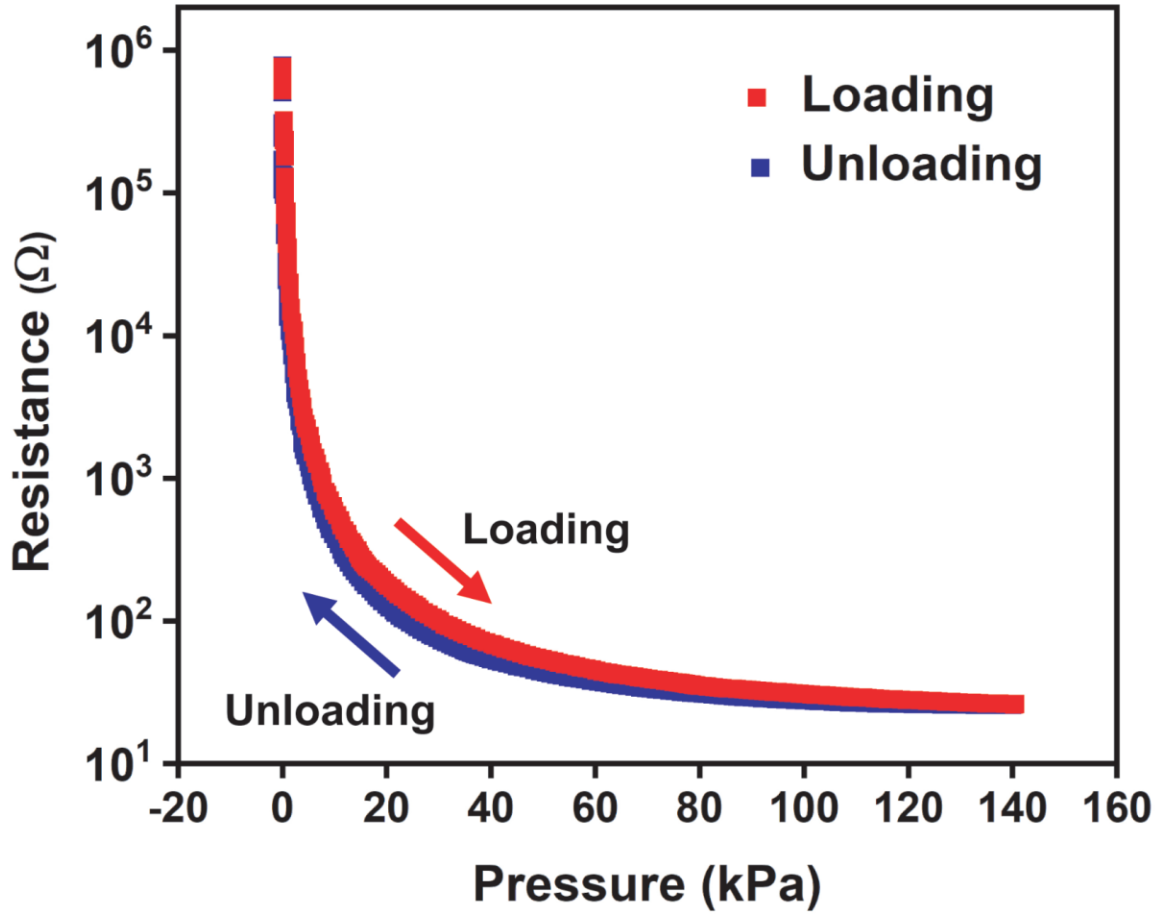
NeuTouch

Pressure response curve of the graphene-based pressure transducer.



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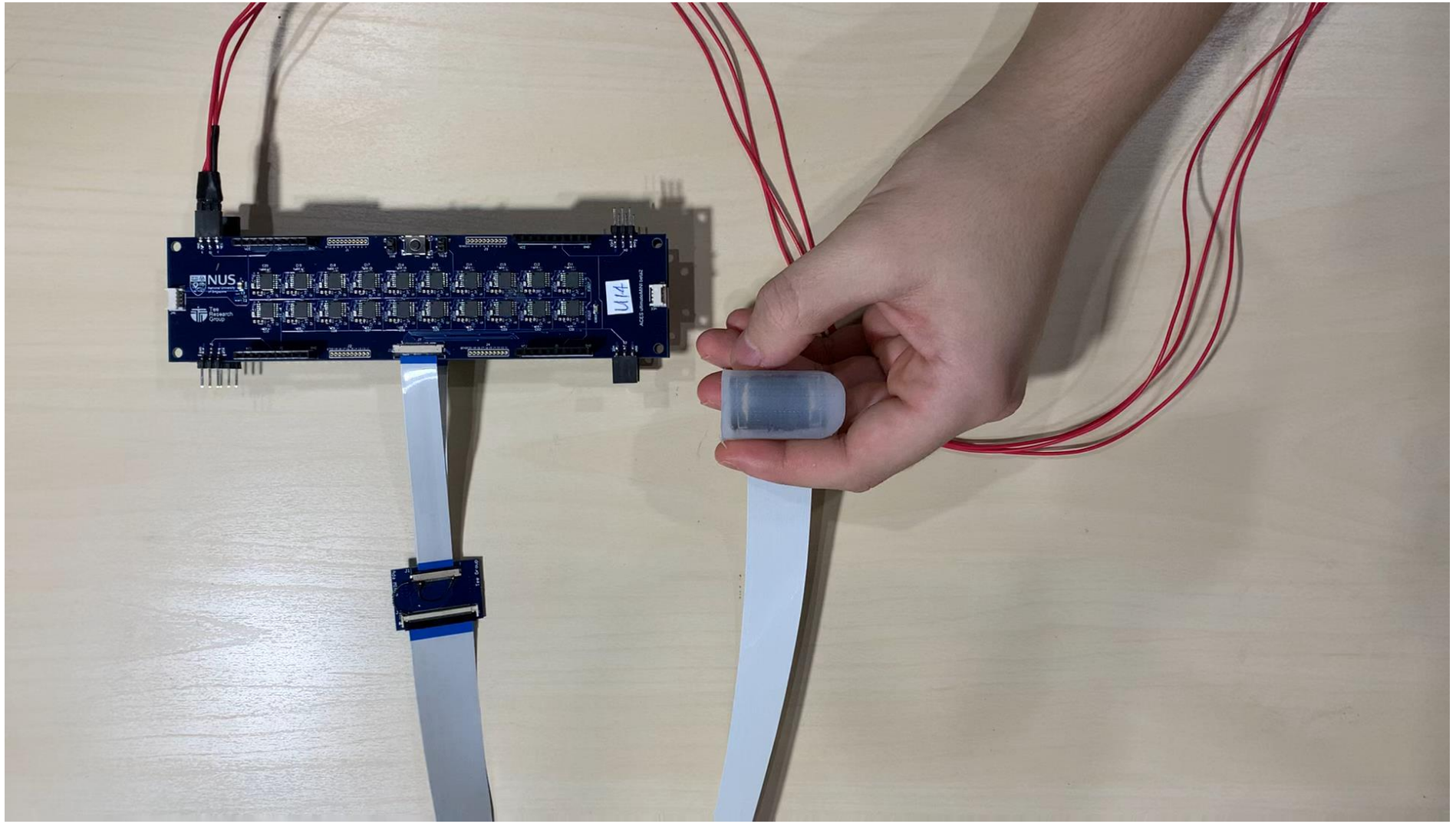
NeuTouch (ACES)



Visual-Tactile Spiking Neural Network (VT-SNN)

Preprocessing:

$$V_w = \begin{cases} 1 & \sum_w S \geq S_{min} \\ 0 & \text{otherwise} \end{cases}$$

Raw spike train

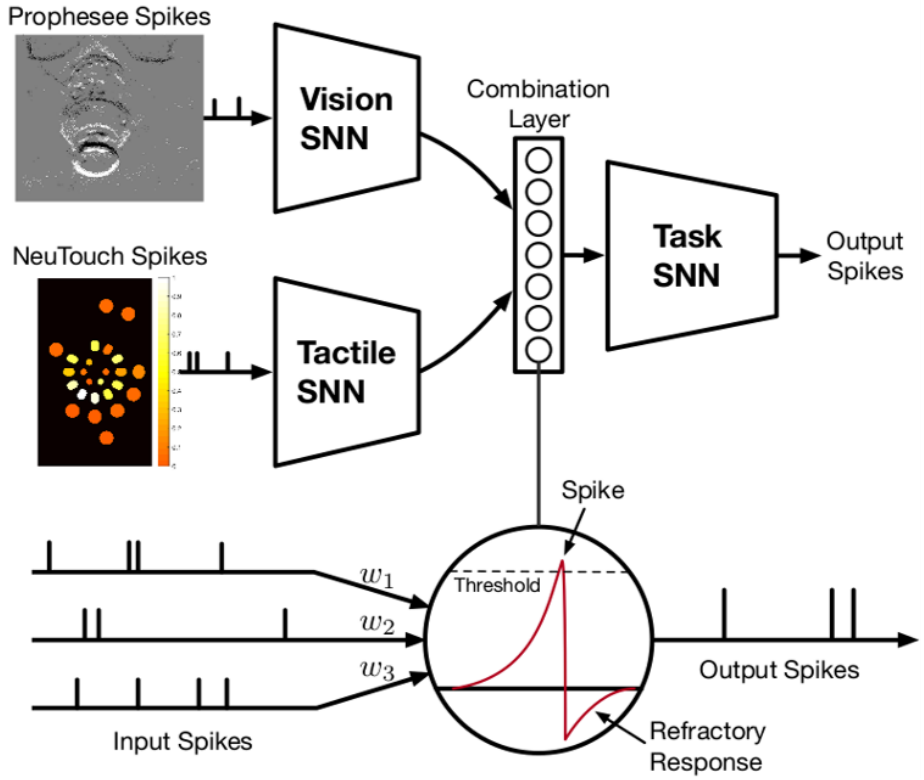
Threshold

Binning window length

Neuron Model (Spike Response Model):

$$u(t) = \sum w_i(\epsilon * s_i)(t) + (\nu * o)(t)$$

An output spike is generated whenever $u(t)$ reaches a predefined threshold ϑ.



Visual-Tactile Spiking Neural Network (VT-SNN)

Training:

Our VT-SNN is trained using SLAYER[1] framework.

Losses

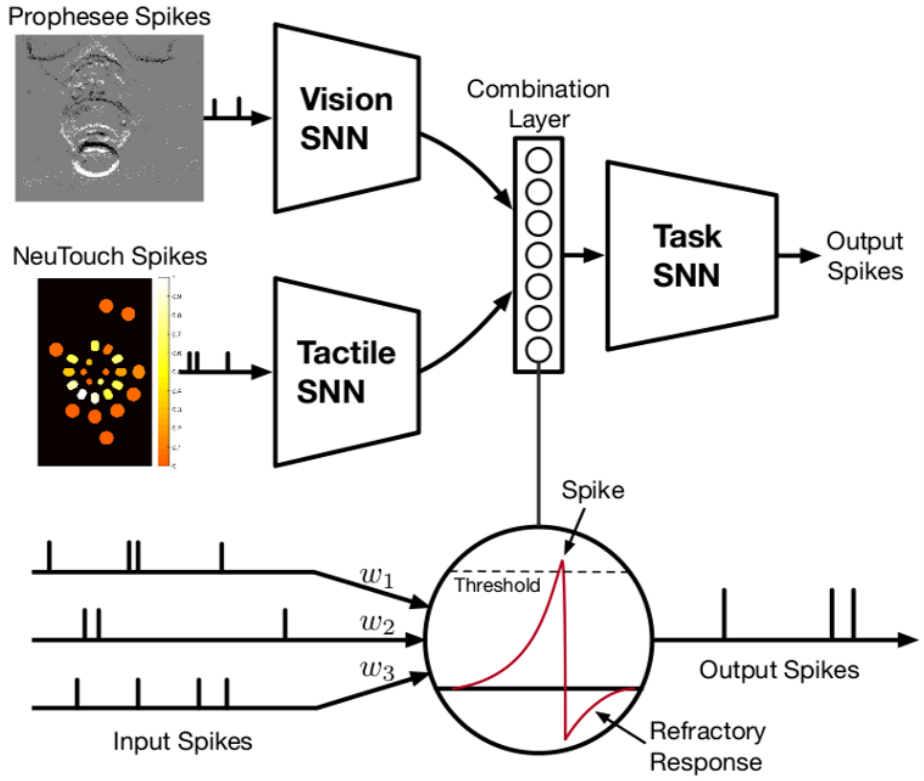
1. Standard Spike Count Loss:

$$L = \frac{1}{2}\sum_{n=0}^{N_o}\left(\sum_{t=0}^{T} s^n(t) - \sum_{t=0}^{T} \tilde{s}^n(t)\right)^2$$

Obtained counts Desired counts

2. Proposed Weighted Spike Count Loss:

$$L_\omega = \frac{1}{2}\sum_{n=0}^{N_o}\left(\sum_{t=0}^{T} \omega(t)s^n(t) - \sum_{t=0}^{T} \omega(t)\tilde{s}^n(t)\right)^2$$



[1]. Shrestha, Sumit Bam, and Garrick Orchard. "Slayer: Spike layer error reassignment in time." *Advances in Neural Information Processing Systems*. 2018.

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Visual-Tactile Spiking Neural Network (VT-SNN)

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We tested VT-SNN framework on two robotic tasks:

- Container and weight classification
- Rotational slip classification

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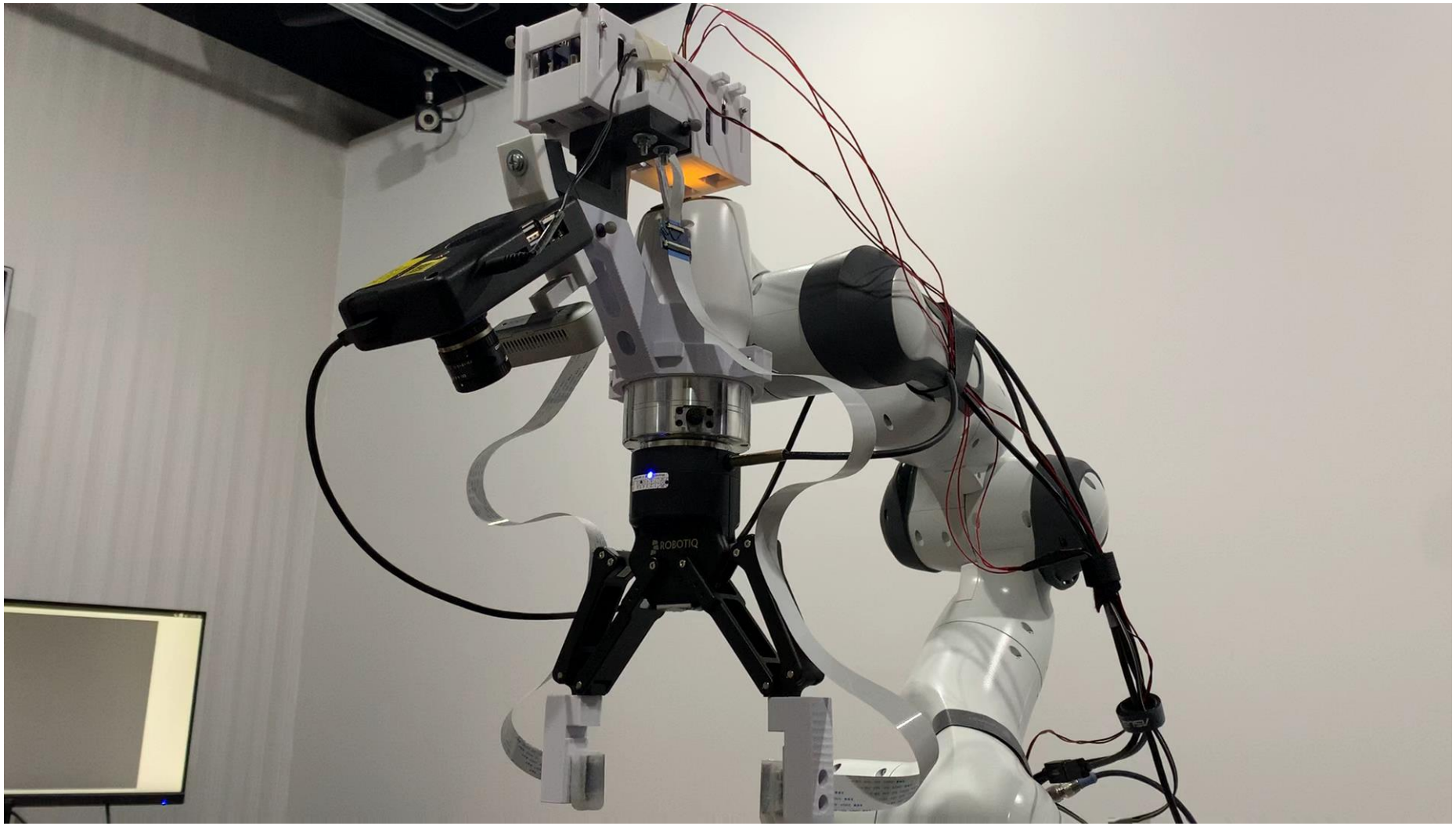
Robotic Setup



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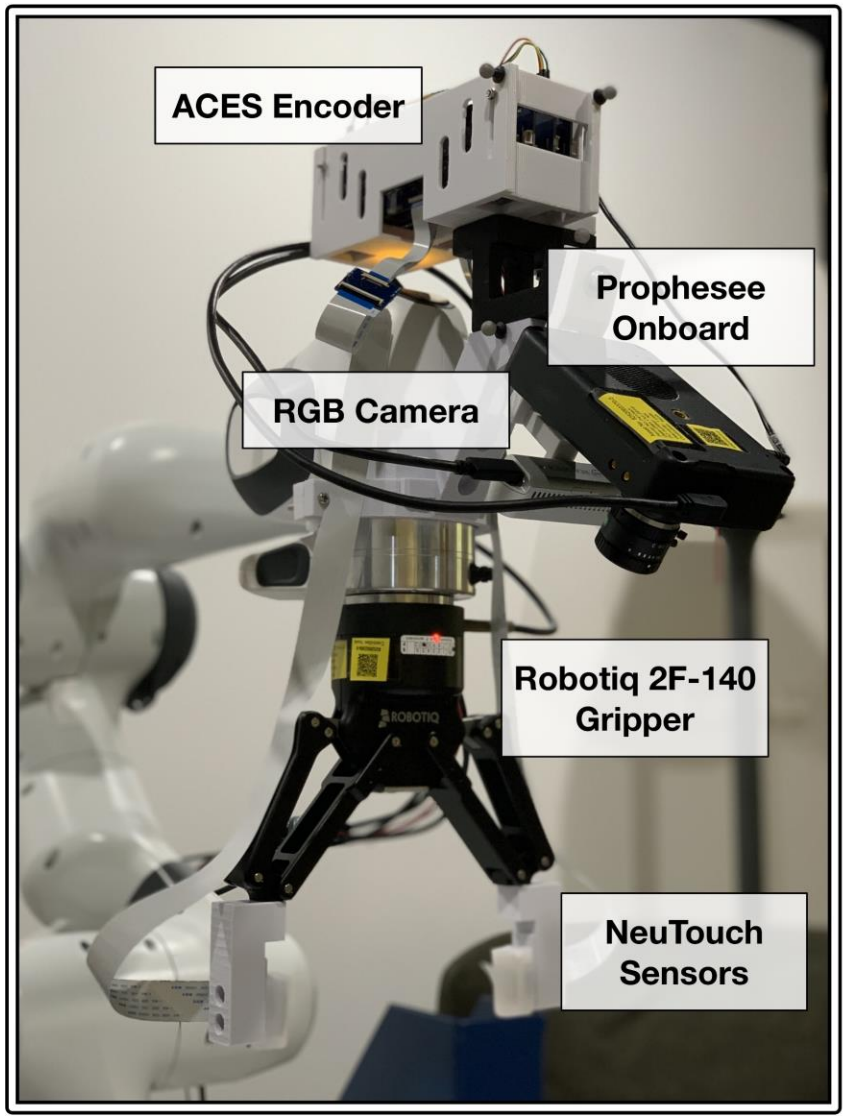
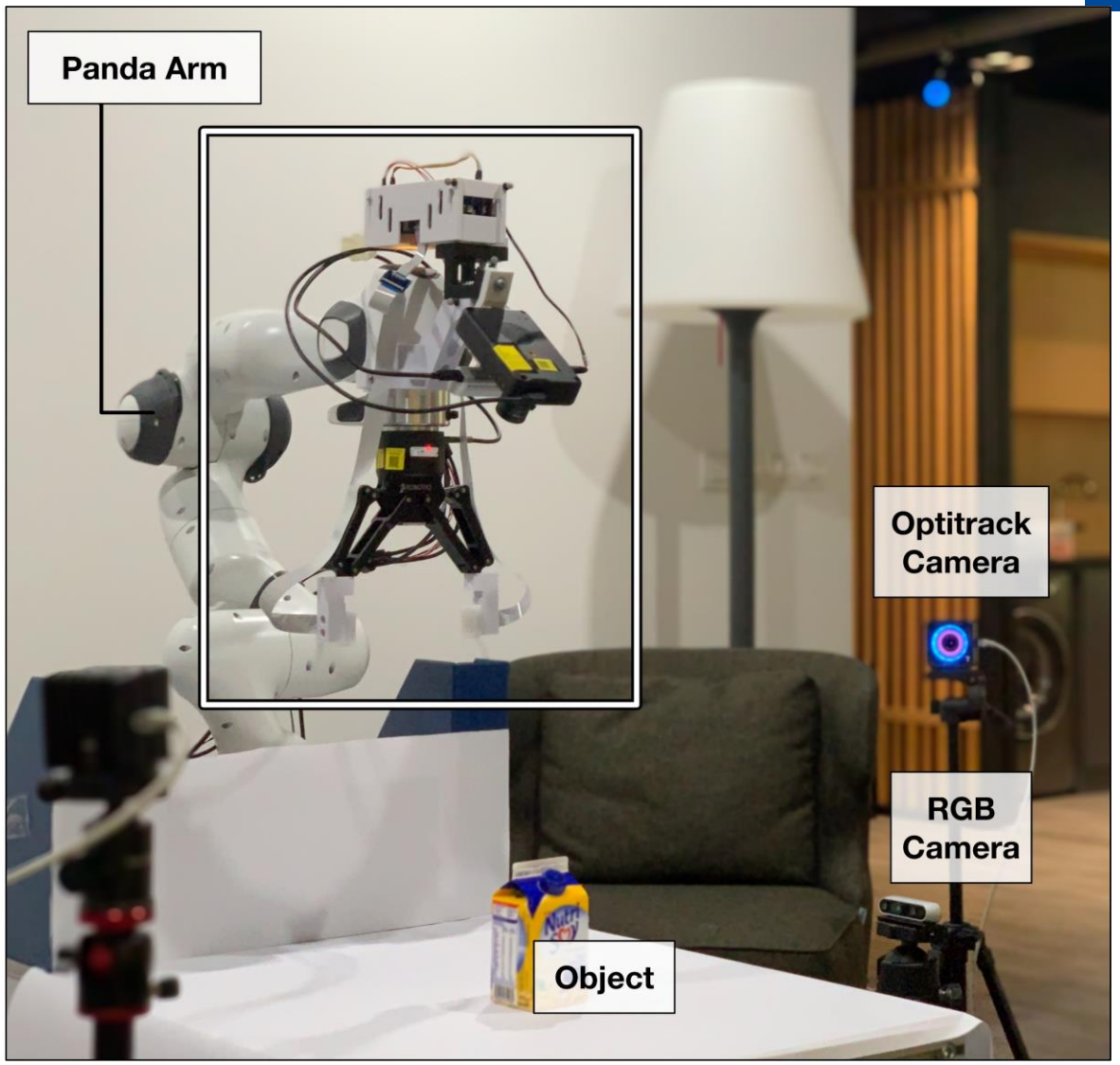
Visual-Tactile Spiking Neural Network (VT-SNN)





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Container and Weight Classification

- 4 objects, 5 different weights = 20 classes
- 15 samples per each class = 300 data points
- 5-fold cross validation for testing the models

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Container and Weight Classification: Data Collection Setup

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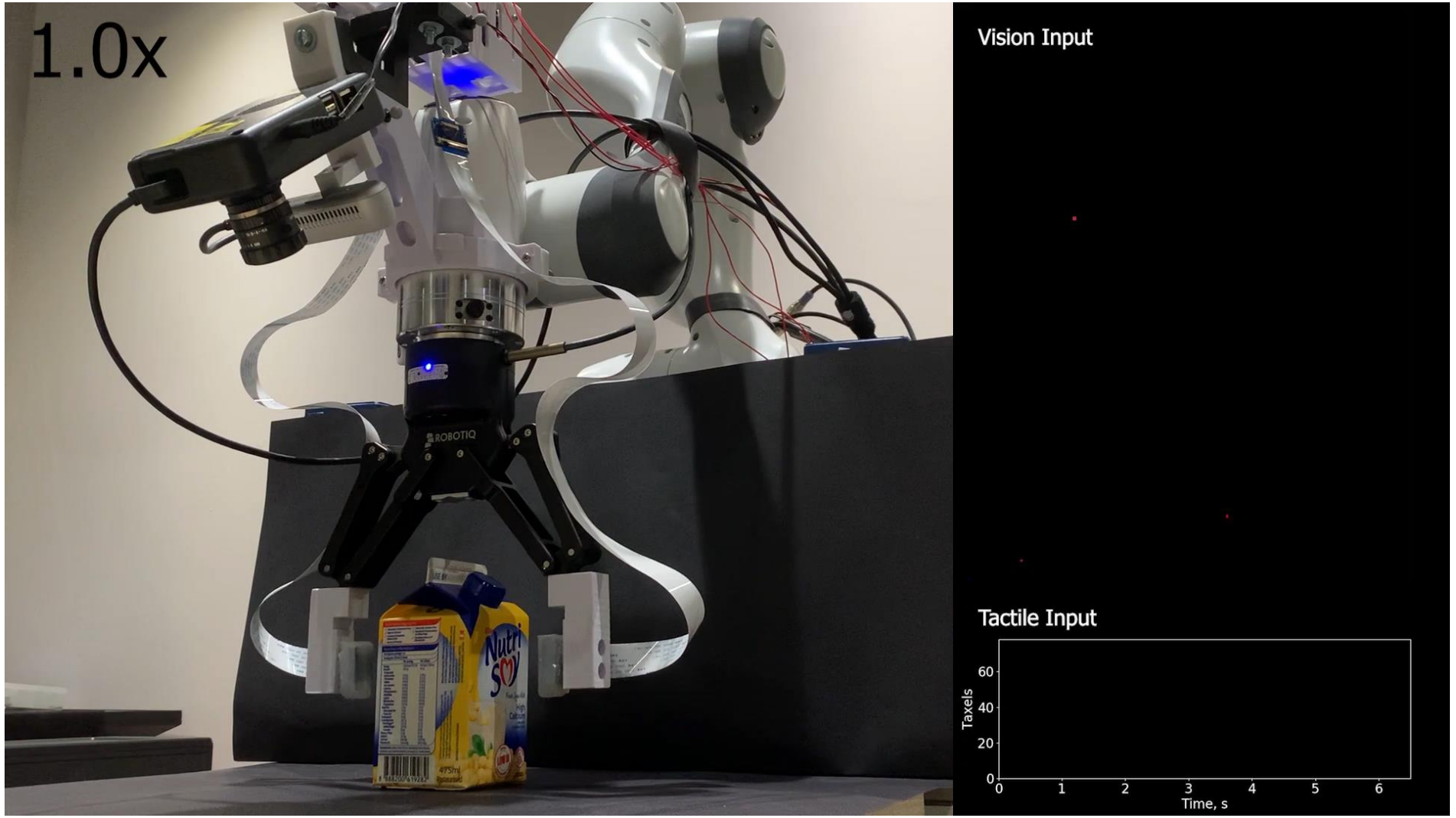

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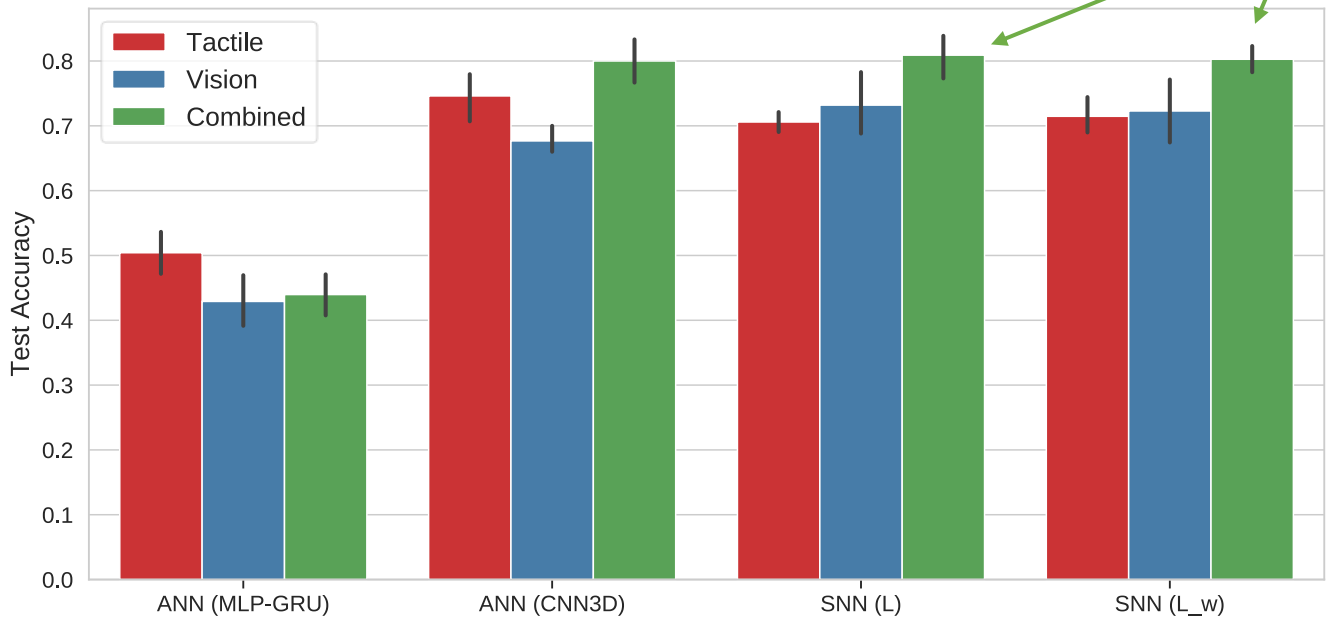
Container and Weight Classification: Results (Final Accuracy)

Multimodal gives higher accuracy



SNN models perform better or comparable to ANN models

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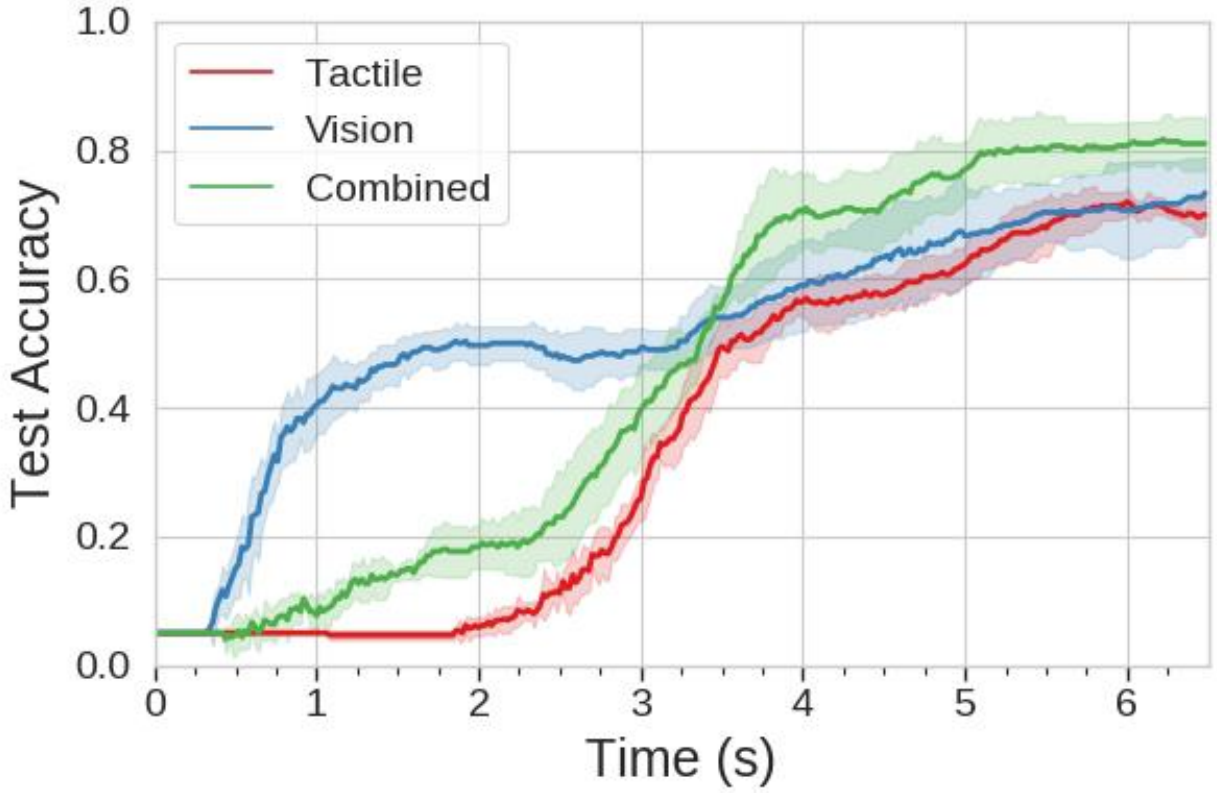
Container and Weight Classification: Results (Early Classification)

$$L = \frac{1}{2}\sum_{n=0}^{N_o}\left(\sum_{t=0}^{T} s^n(t) - \sum_{t=0}^{T} \tilde{s}^n(t)\right)^2$$



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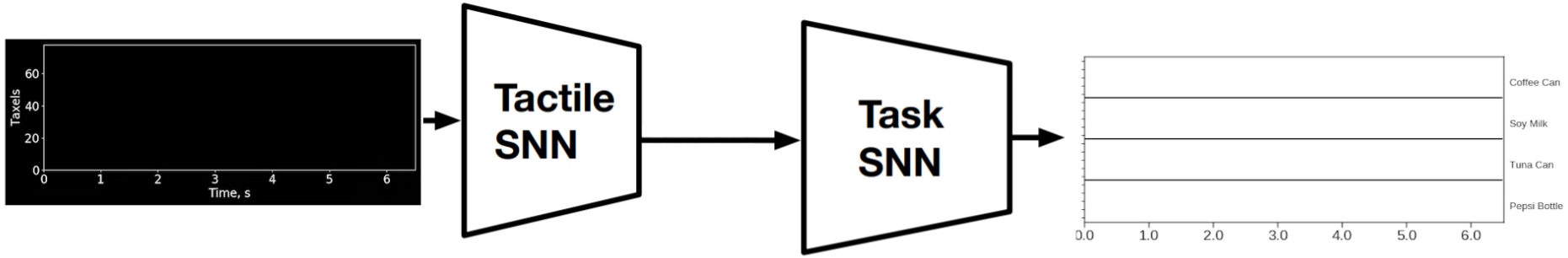
Container and Weight Classification: Output Example



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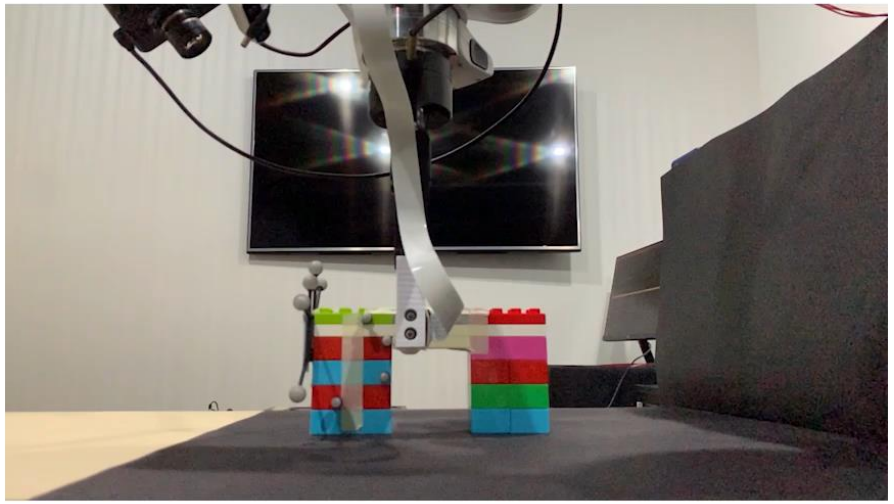
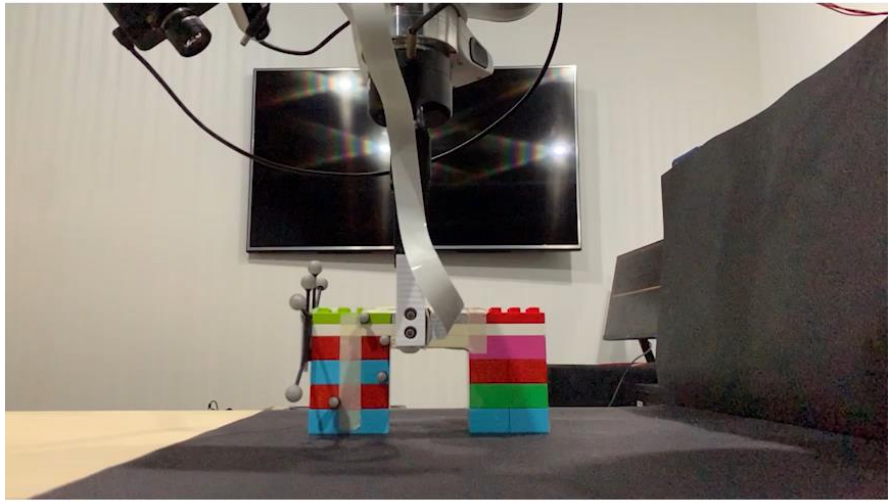
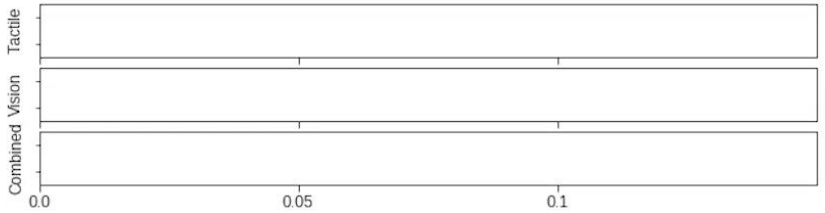
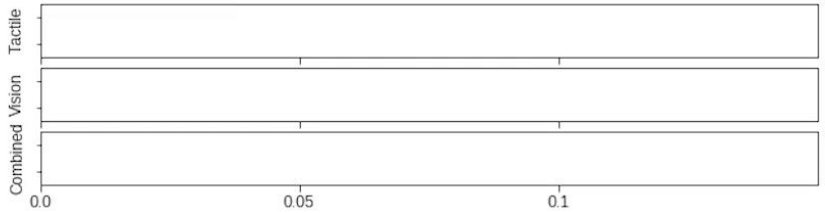
Rotational Slip Detection: Example of Output Neuron

0.02x









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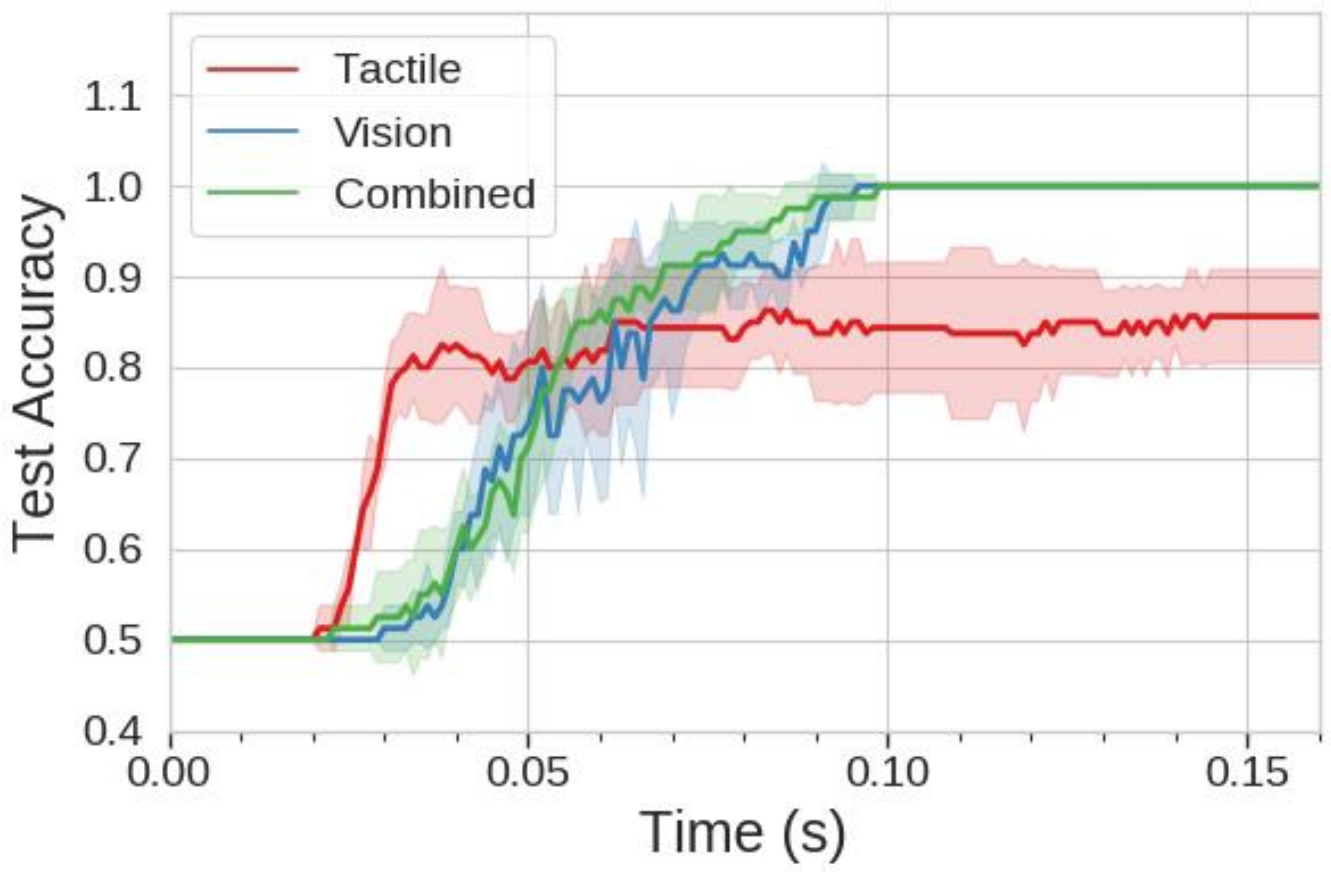
Rotational Slip Detection: Results (Early Classification)

$$L = \frac{1}{2}\sum_{n=0}^{N_o}\left(\sum_{t=0}^{T} s^n(t) - \sum_{t=0}^{T} \tilde{s}^n(t)\right)^2$$



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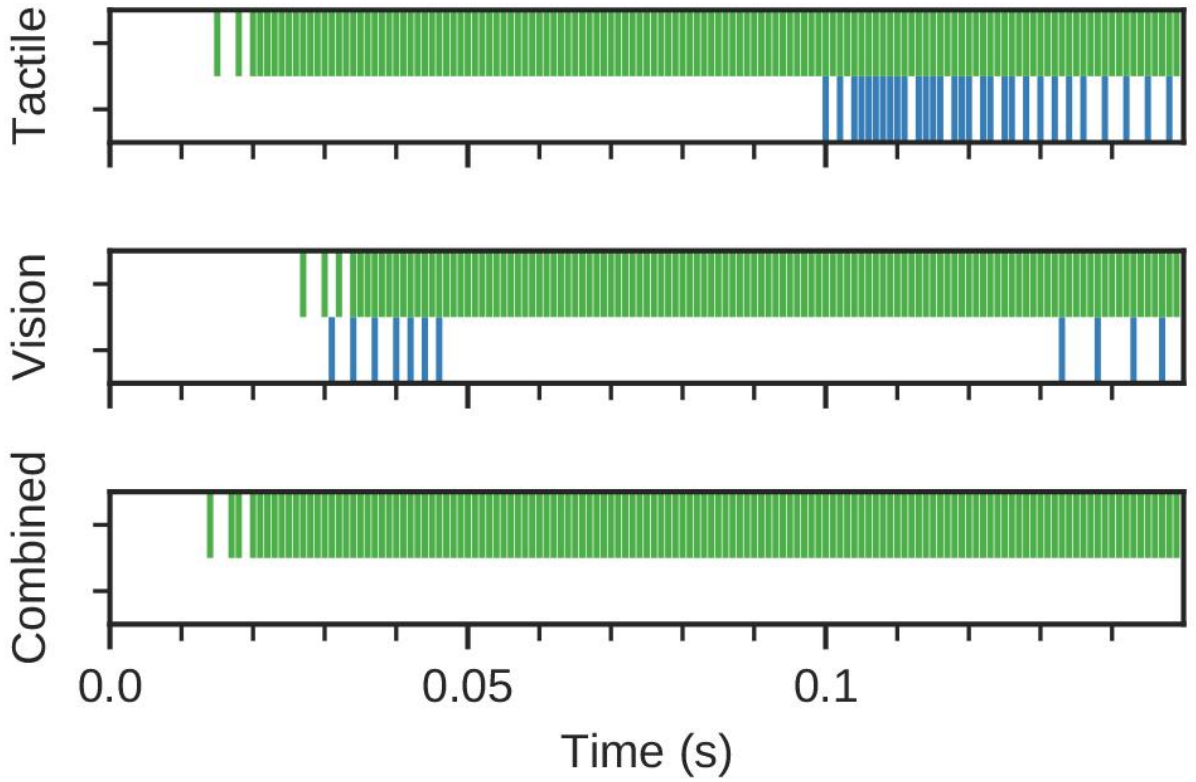
Rotational Slip Detection: Output Example with Weight Spike Count Loss

$$L_\omega = \frac{1}{2}\sum_{n=0}^{N_o}\left(\sum_{t=0}^{T}\omega(t)s^n(t) - \sum_{t=0}^{T}\omega(t)\tilde{s}^n(t)\right)^2$$



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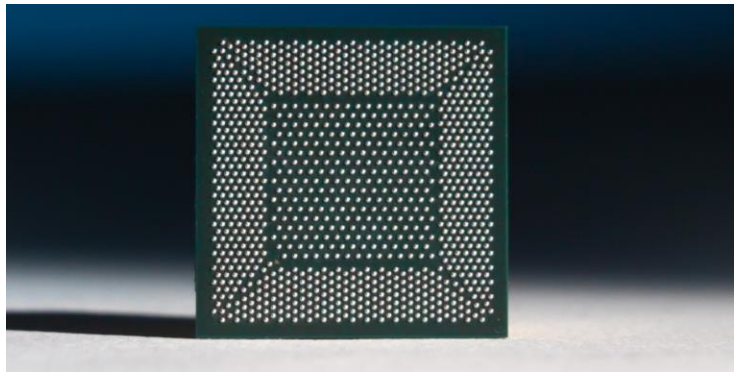
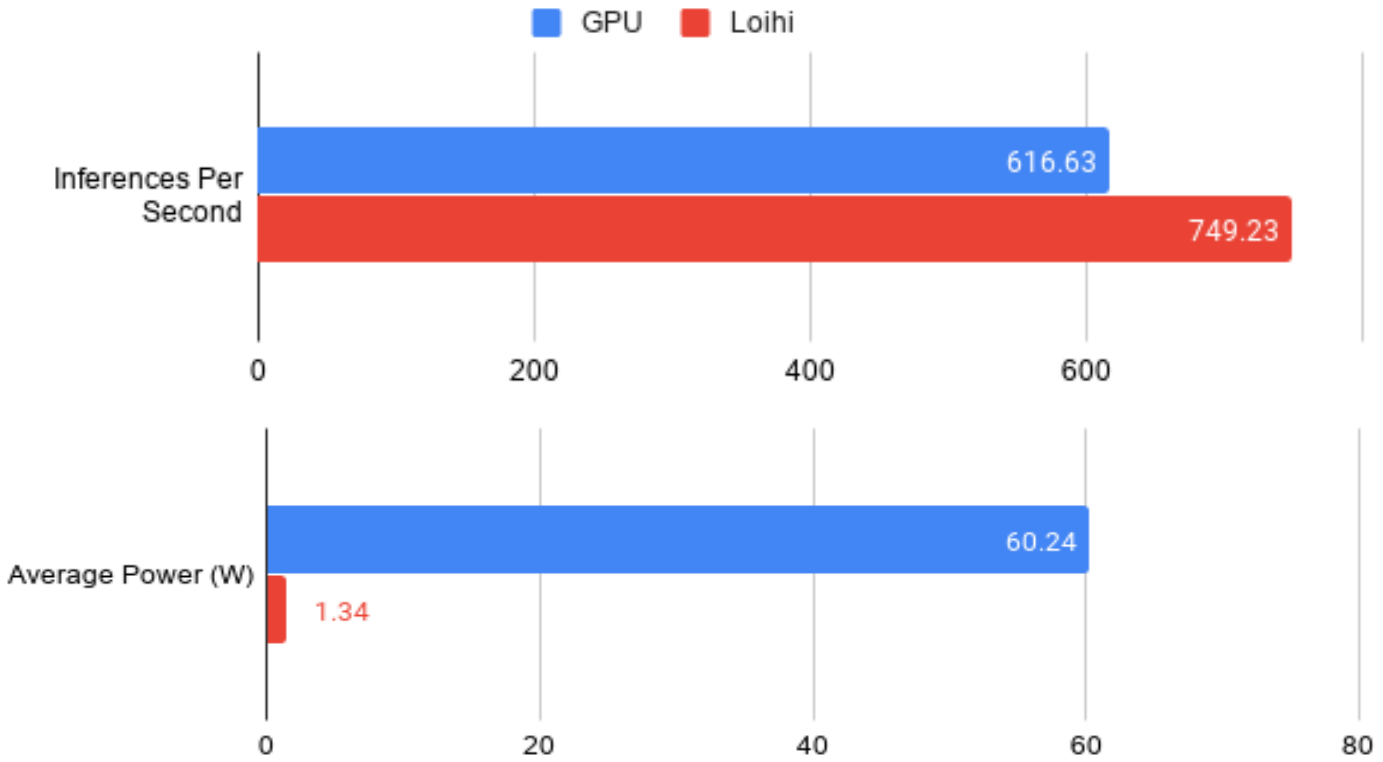
Power Utilization

Nvidia RTX 2080 Ti



Intel Loihi



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Conclusion

We proposed

- **NeuTouch** – event-based tactile sensor.
- **A VT-SNN** framework that combines side and touch event-based data.
- VT-SNN framework is tested on **two robot tasks**:
  1. Container and weight classification
  2. Rotational slip classification
- We put our datasets/code publicly available and can be found with following link:

https://clear-nus.github.io/visuotactile/