### Deep Learning On Embedded Computers

#### Adding Smart Sensors To Your Data Collection Pipeline

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#### HASAN A. POONAWALA

poonawalalab.github.io **Robotics & Automation** left wall •  $p_{\rm max}$ Motion planning and •  $\theta \downarrow -p$ control Dynamics & Control right wall Machine Learning for • (Robotics) LiDAR navigation control Robotic manipulation **Biological control** • Our Work systems: spiking neural networks Optimization Machine Learning (Inc. Traj Opt, (Perception) MPC, Planning) **Sponsors:** • Al-enabled machining Small Sat EM control Physics-based ML for Additive Manufacturing ARORATORY and Thermofluid sim University of Kentucky. Mechanical and Aerospace Engineering 1

## Outline

- Embedded Computers
- Edge ML Applications
  - Al Coach Device
  - Object-aware Robot Navigation
  - Perception for Robot Arm Control
- Non-NVIDIA Microprocessors

## **Embedded Computers**

An embedded system is a computer system built into a larger mechanical or electronic device to perform a specific function.

#### Examples





#### Specs

#### Intel Core i9 (PC) STM32





Cores	18	1 or 2	
Address length	64 Bits	32 bits	
Cache memory	32 MB	32 KB	
RAM	16 GB (ext)	1 MB	
NV Memory	1 TB (ext)	2 MB	
Clock Speed	6.0 GHz	600 MHz	
Cost	\$300-\$500	\$8-\$20	

#### **Embedded GPU Specs**



#### **Inference Time**



Deep Learning on Embedded Computers. CCS/ITSRCI Seminar Series on AI in Practice

#### Accuracy



Source: MulTiNet: Multimodal Neural Networks for Glaucoma Based on Transfer Learning

## Al Coach

## Background

• 2020 UK Igniting Research Grants pitch:

Use new GPU-enabled embedded systems to provide realtime feedback on body movement



### Background





#### **TRT-Pose**



Demo



Jetson Nano

- TRT: TensorRT, NVIDIA's high-performance deep learning inference library
- TRT Pose: optimized deep learning model for real-time pose estimation on NVIDIA Jetson devices
- 80 MB Model

#### **TRT Pose**



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#### **Research Goal**



Observe study participants in their home while they work at a desk.

#### **Al Coach Device**





#### Sit vs Stand vs DK Classification



#### Example Data 1

Туре	Duration (minutes)
Total	480
Seated	360
Standing	5
Not visible	115



#### Example Data 2

Туре	Duration (minutes)
Total	75
Seated	14
Standing	17
Not visible	44



## Highlights

**Extended remote data collection** 

Collected data on a ~\$200 device over weeks in participant's homes

#### No Wi-Fi

Avoids privacy and cybersecurity concerns (and complex IRB)

**Better Models/Devices** 

AI Models and embedded ML devices have advanced a lot since 2021.

#### **DL for Smart Sensors and Data**









# Tracking Specific Objects

#### **Tracking Robot Arms**









#### **DREAM Model**



https://sim2realai.github.io/assets/img/2020-06-29/Lee\_etal\_2020\_dream\_panda\_reaching\_frame.png

#### **DREAM Model**



https://sim2realai.github.io/assets/img/2020-06-29/Lee\_etal\_2020\_dream\_pipeline.png

 80 - 200 MB model depending on the backbone (VGG-X vs ResNet-X)

#### **DREAM On Jetson Nanos**



### **Tracking Manipulation Objects**



Image credit: Varun Hariprasad

# Object-aware Navigation

### **Navigation using Deep Learning**



#### **End-to-end** Neural Net Control



### **Example: Aircraft Taxiing**

#### Image-based Neural Network Controllers

Image-based neural network controllers will allow for low-cost and flexible control systems in a variety of applications.



#### Source: slides by Sydney Katz

#### Navigation at UK



- Simple linear controller < 1 MB size
- Guaranteed to navigate hallway without crashing

#### Results



#### **Distance-based Navigation**



#### **Object-aware Navigation**



#### **Yolov9 Segmentation Model**



Source: https://img.freepik.com/premium-vector/vector-realistic-washing-machine-white-3d-mockup\_208581-782.jpg?semt=ais\_hybrid



## Non-NVIDIA Microprocessors

#### Examples

	STM32N6	ESP32s3	
CPU	Arm <sup>®</sup> Cortex M55 (1)	Xtensa <sup>®</sup> LX7 (4)	
Memory	4.2 MB	2 MB	
Cores	300 MAU <sup>1</sup>	-	
Cost	\$8-\$20	\$2 - \$10	
Speed	0.6 TOPS	-	
Clock Speed	1 GHz	240 MHz	

#### **Performance: STM32N6**

#### Inference per second



### Performance: ESP32s3

- Custom CNN 500 KB
- Inference in ~60 milliseconds
- 93% Accuracy



### **Single Instruction Multiple Data**





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#### **Thank You!**





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