

2025-2026 Smart Prosthetics PDR



ME 486A Peter Bishay

9th Cohort Team Members:

Abdulrahman Al Dulaimi, Jorge Alvarez, Justin Brown, Cesar Flores Velasquez, Nathan Handal, Chris Lopez, Vanessa Montejano, Hao Phan, Matthew Portillo, Sergio Rivas, Kainen Shaw, Yaroslav Tretiak, Kurt Trocino, Katsuki Yasuda



Agenda

- Problem Statement
- Arm Design
 - History
 - Adaptive Attachment system
 - Current arm design
- Foot Controller System
 - History
 - Current design
 - Neural networks
- Haptic Feedback
 - History
 - M.I.D.A.S. Haptic system
- Q&A



Problem Statement

- Most prosthetics are made for amputees and can't be used by people with temporary arm or shoulder injuries.
- They are often expensive and not ideal for short-term use or rehab.
- They often rely on EMG-control which can be unreliable due to skin or muscle issues.
- Few provide real sensory feedback, and those that do give limited response.
 - Vibrotactile (vibrations)
 - Electrotactile (electricity)

Arm Design

Team members:

Abdulrahman Al Dulaimi, Justin Brown,
Cesar Flores Velasquez, Sergio Rivas,
Yaroslav Tretiak, Katsuki Yasuda

Goals & Objectives

- Create a lightweight user friendly modular robotic arm.
 - Wearable Third Arm
 - Shoulder-and-Below Prosthetic
- Fast attach/swap: one-handed, tool-free connections
- Safe & reliable
- Range of motion:
 - 1 DOF in the shoulder and elbow
 - 2 DOF in the wrist
- Weight of complete system ~2.2 kg (5 lb)
- Finger Grip Strength ~60 N

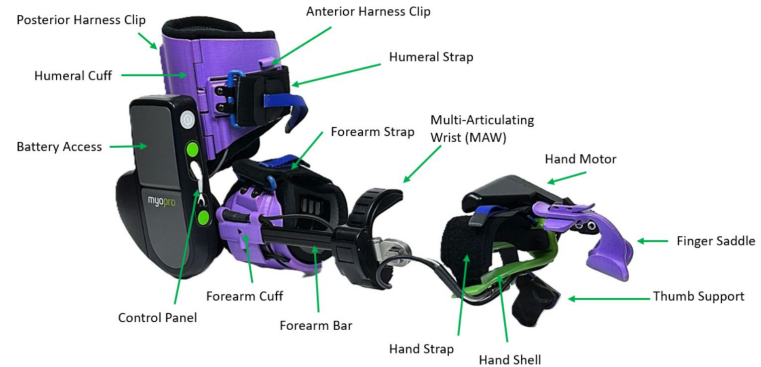


What is currently on the market?



Kinova Assistive: Jaco Arm

- Third arm system targeted for wheelchair users
- Controlled with joystick



MyoPro

- “designed to help restore function to the wearer’s paralyzed or weakened upper extremities”

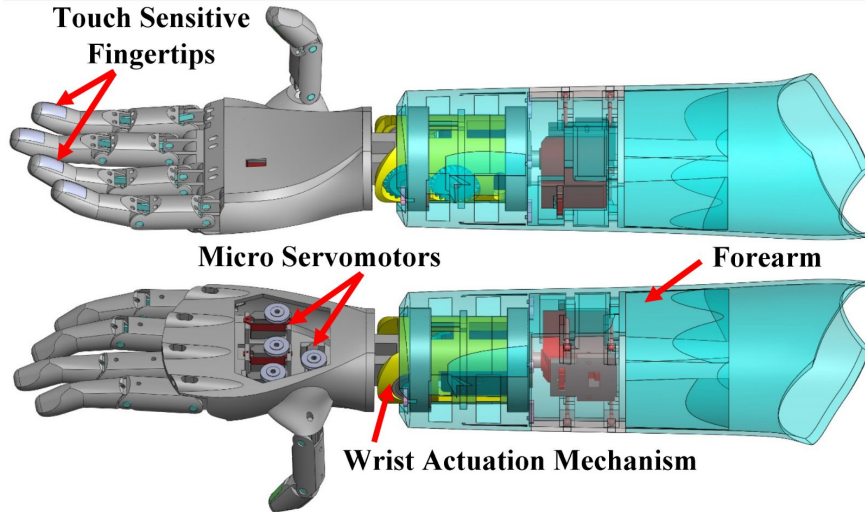
Other Third Arm Systems

Jizai Arms

- Focused for Aesthetics
- Attach & Detach in multiple mounting points
- Controlled by external user

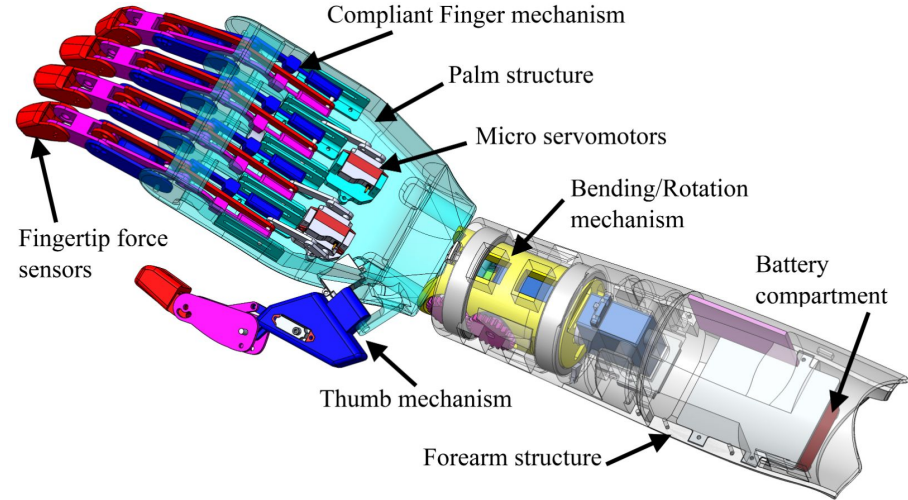


5th and 6th Cohort Arms



5th Cohort

- 1 DOF Fingers & Thumb
- Tendon Base

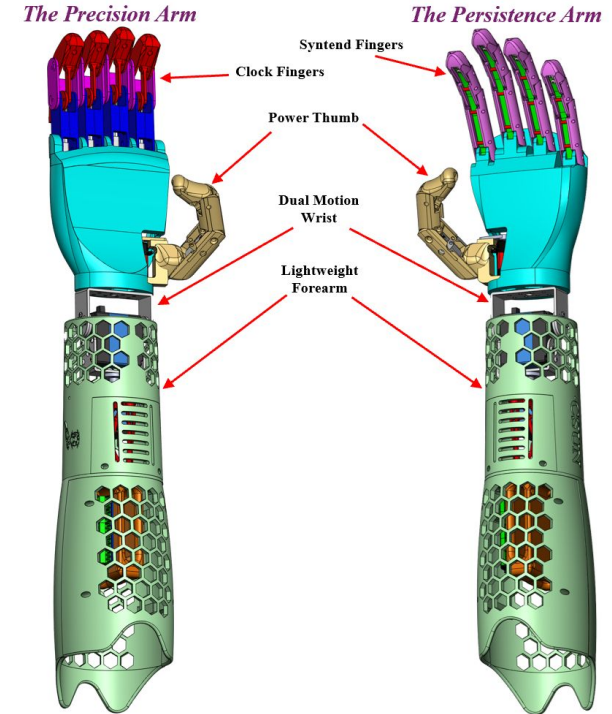


6th Cohort

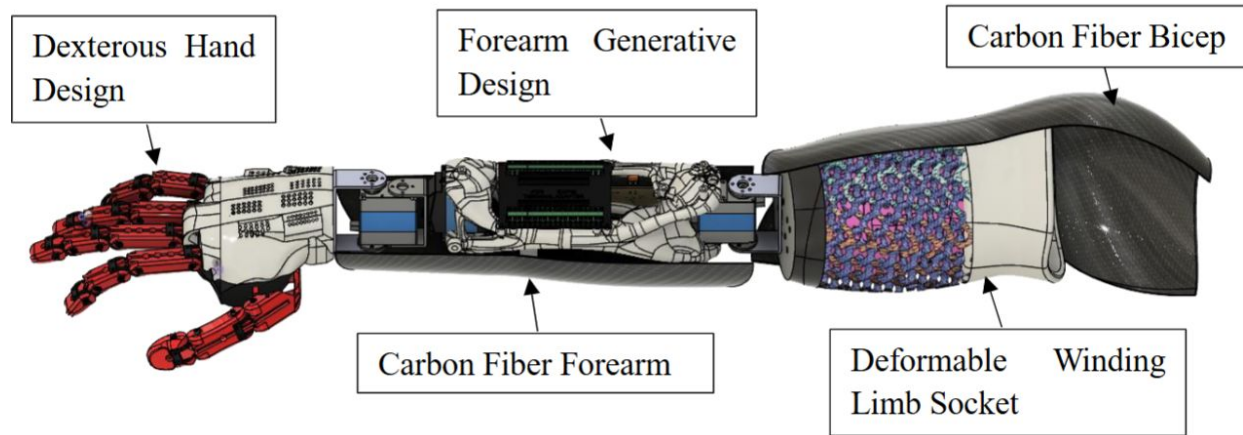
- 1 DOF Fingers, 2 DOF Thumb
- Link Based

7th Cohort “Precision” & “Persistence” Arm

- Elbow-and-Below Prosthetic
- Precision Arm: **Link** based joints
- Persistence Arm: **Tendon** based joints
- 1 DOF Fingers
- 2 DOF Thumb
- Fusion’s **Generative Design**

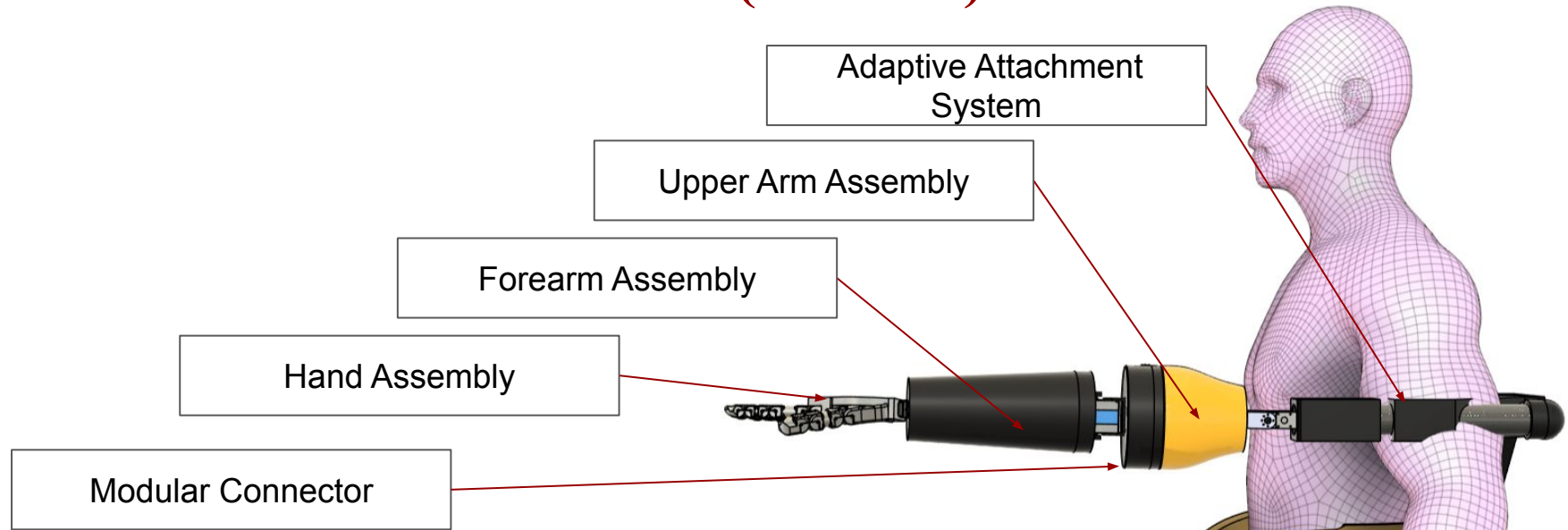


8th Cohort “Prometheus” Arm



- Shoulder-and-Below Prosthetic
- **Tendon** based fingers with **Rolling Contact Joints**
- 2 DOF Fingers & Thumb
- **Generative Design**

9th Cohort “Cerberus” Arm (Current)



- Tendon base fingers with rolling contact joints
- 2 DOF Fingers & Thumb
- Attachment system for comfortable, stable, and modular wearability.

The Adaptive Attachment System allows the wearable robotic arm to be used in various ways such as



*Figure showing the Kinova Assistive:
Jaco Arm*



*Figure showing robot arm on the Robot
fixture table from Cabot Depot*



*Figure showing a hydraulic wearable
arm from Université de Sherbrooke in
Canada*

<https://assistive.kinovarobotics.com/product/jaco-robotic-arm>
<https://spectrum.ieee.org/robotic-third-arm-can-smash-through-walls>

What is the Adaptive Attachment System

- Allows robotic arm to **quickly connect** to different mounts or tools.

- The locking mechanism keeps the arm **stable** and **prevents accidental release**

- Designed so users can attach or detach easily, even with **limited dexterity**.

- Users can **switch attachments** or mounting points in seconds, without tools



Adaptive Attachment System

Version 1 of the Adaptive Attachment System

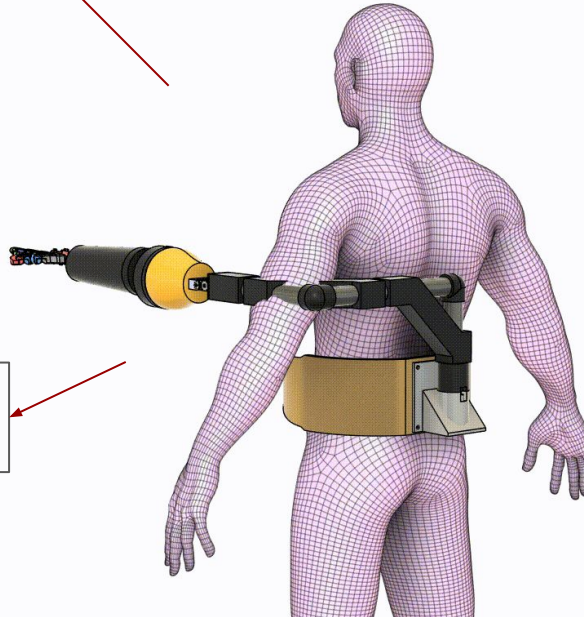
Functions as the natural extension of the user

On-the-hip belt quick attachment system

Carbon fiber shoulder quick attach system

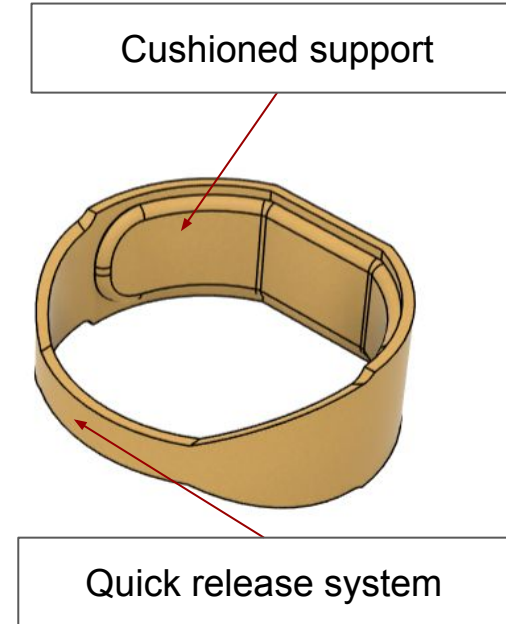
On the hip wearable design

Quickly connect and disconnect robot arm from system

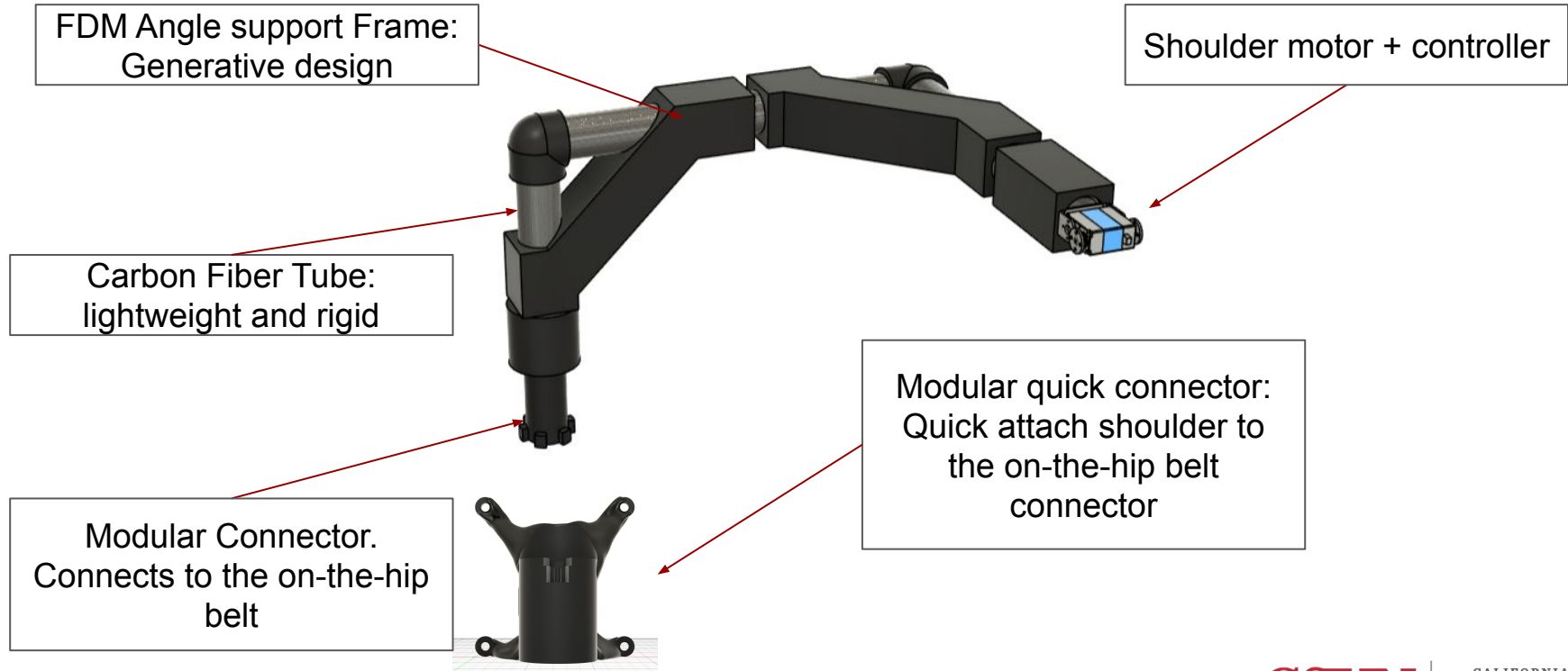


On-the-hip Belt Connector

- Belt mounted version of the Adaptix attachment system
- Main support is worn around the waist reducing strain on the upper body of the user.
- Must be easy to attached and detach from user with one arm and minimal dexterity
- Acts as the foundation for the shoulder assembly that connects the arm to the Adaptix system

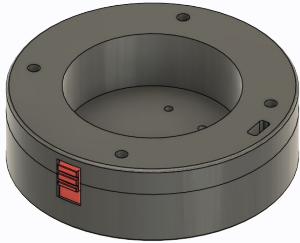


The shoulder

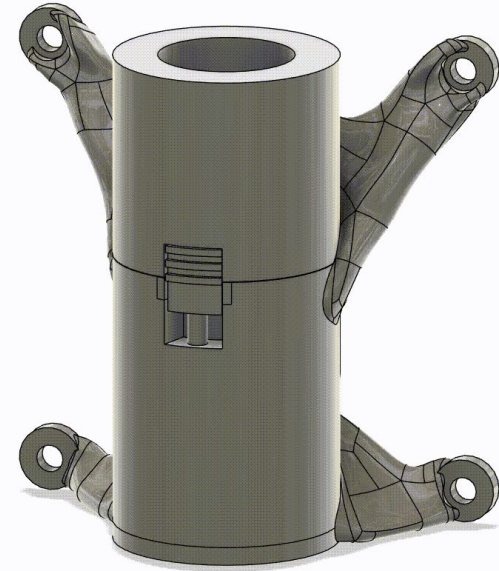


Modular Connector

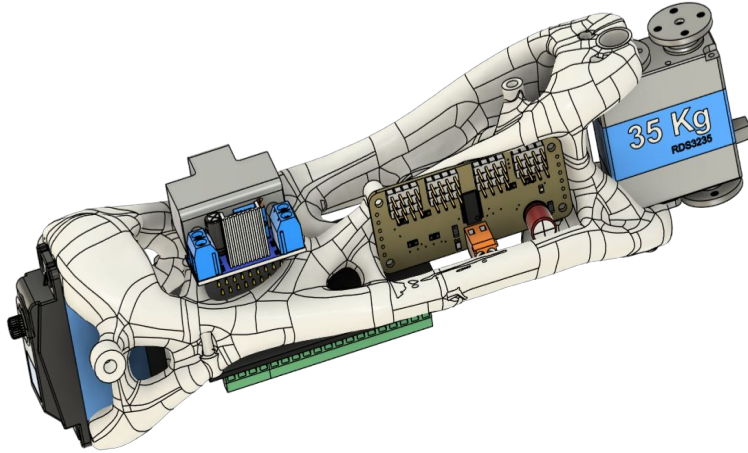
- Enables easy attachment and detachment of the prosthetic arm in multiple locations.
- Smart Connector: made for the bicep, has magnetic link wires to allow wires to attach and detach.



Smart Connector



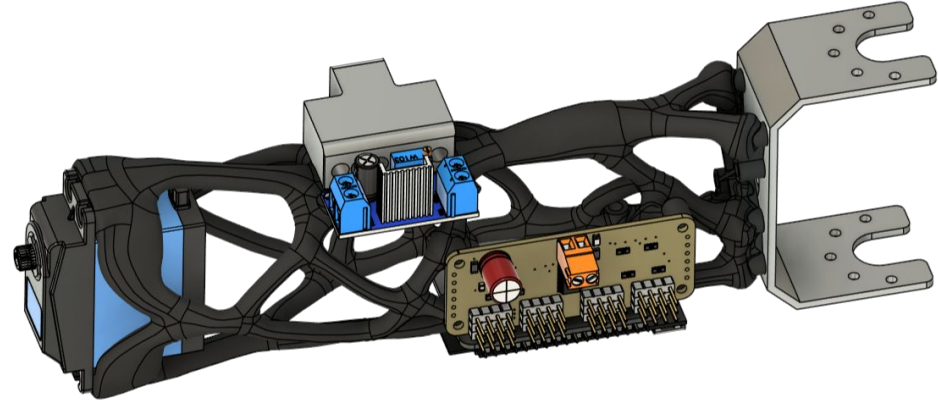
Forearm



8th cohort Generative design forearm

Net Weight: 134 grams

Material : ABS



Current Generative design forearm

Net Weight: 57 grams

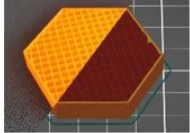
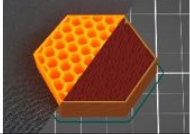
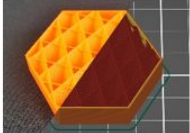
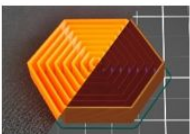
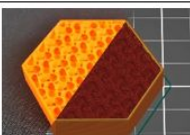
Material : 20 % Carbon Fiber Reinforced Nylon 6

Forearm Material Selection

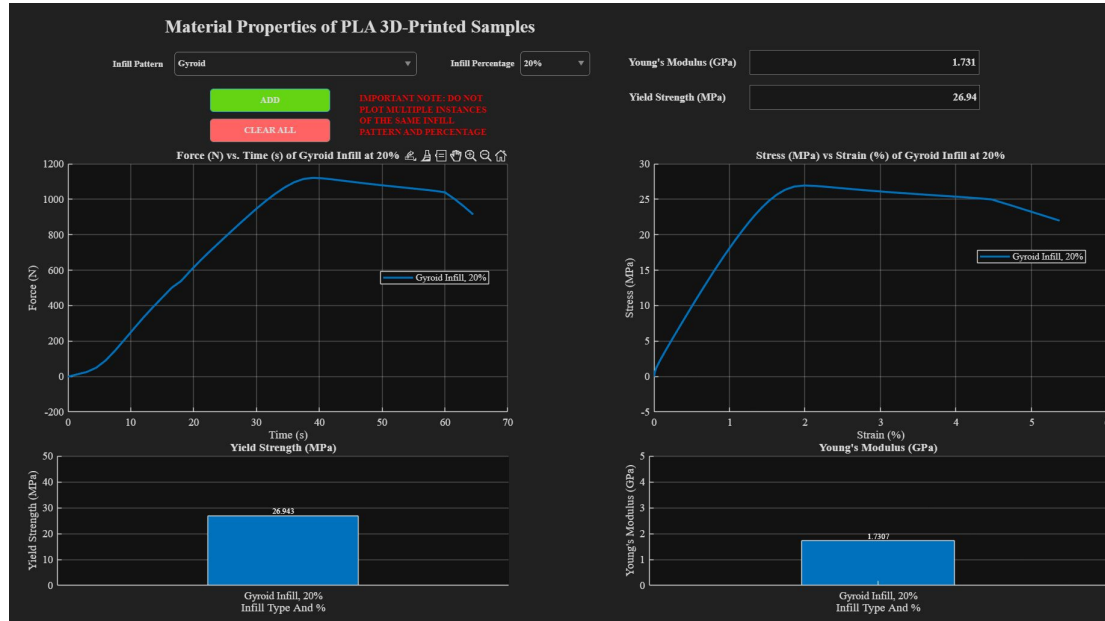
Property	ABS	PC (Polycarbonate)	Fiberon PA6-CF20
Strength & Stiffness	Moderate (~40 MPa, 2 GPa)	High (~60 MPa, 2.4 GPa)	Very high (~110 MPa, 8.6 GPa)
Weight / Density	Light (1.05 g/cm ³)	Moderate (1.20 g/cm ³)	Light-moderate (1.17 g/cm ³)
Heat Resistance (HDT)	~100 °C	~115 °C	~215 °C
Durability	Good impact, moderate rigidity	Excellent impact, moderate rigidity	Very stiff, lower impact flexibility
Printability	Easy; moderate warp	Difficult; needs enclosure	Moderate; needs dry filament & hardened nozzle
Best Use Case	Basic prototypes	Tough shells needing impact strength	High-stiffness lightweight structural parts

20% Carbon Fiber Reinforced Nylon 6 (PA6-CF20) will be used due to its high stiffness-to-weight ratio along with its professional-grade mechanical performance great for load bearing components.

Forearm Material Selection

Infill Type	Characteristics	Advantages	Limitations	
Rectilinear	• Straight intersecting lines	• Simple and fast	• Stress concentrates at intersections	
Honeycomb	• Hexagonal cells	• Efficient load spreading in 2D plane	• Limited Z-direction strength • Longer print time	
Cubic	• 3D lattice of cubes	• Balanced x, y, z rigidity with easy slicing	• Higher material use	
Concentric	• Rings following part perimeter	• Excellent outer wall bonding	• Poor internal rigidity	
Gyroid	• Continuous 3D periodic surface with no sharp corners	• Isotropic mechanical strength in all directions • High stiffness-to-weight ratio	• Relative higher computation for slicing	

Forearm Material Selection

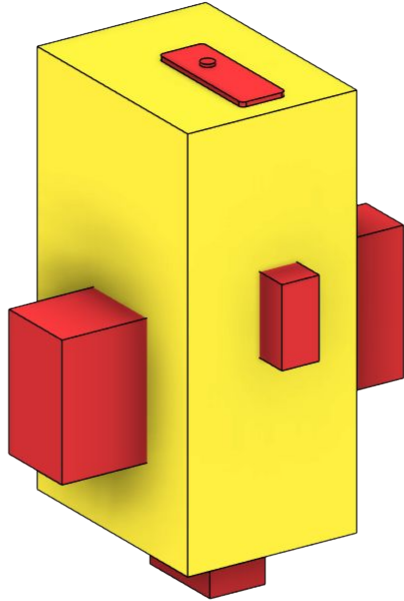


MatLab App Results

Material	Printed PLA 20% Gyroid
Density	1.14E-06 kg /mm ³
Young's Modulus	1.731 Gpa
Poisson's Ratio	0.36
Yield Strength	26.94 MPa
Ultimate Tensile Strength	28.10 MPa
Thermal Conductivity	1.600E-04 W/(mm C)
Thermal Expansion Coefficient	8.570E-05 / C
Specific Heat	1800.00 J/(Kg C)

Fusion Material Properties

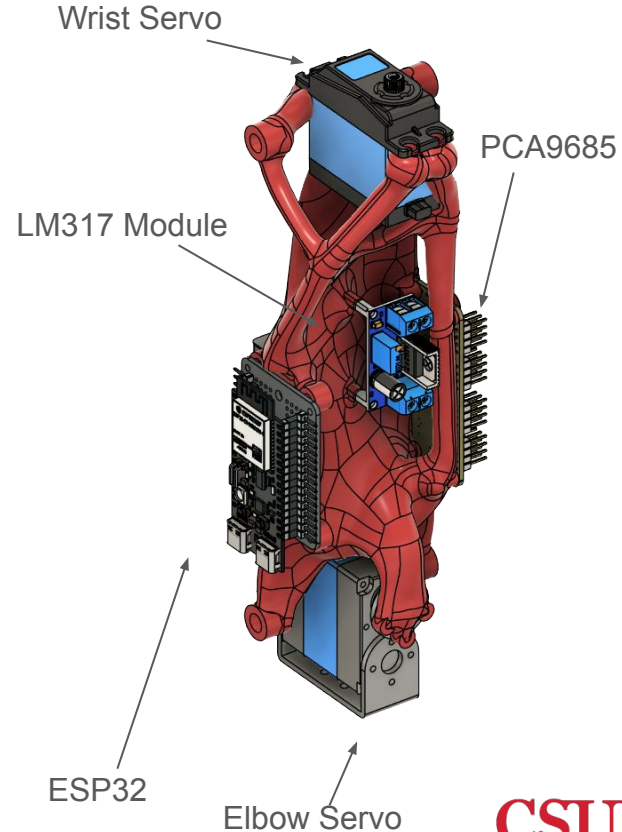
Forearm Material Selection



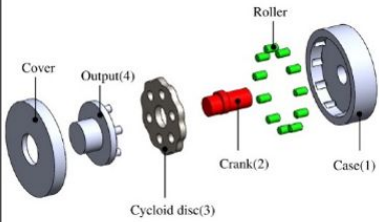
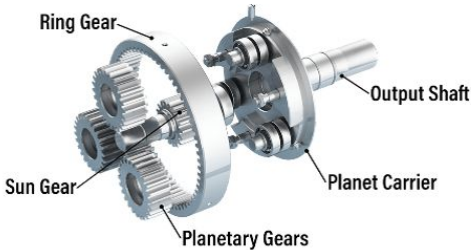
Obstacle and
Starting Shape



Stress View

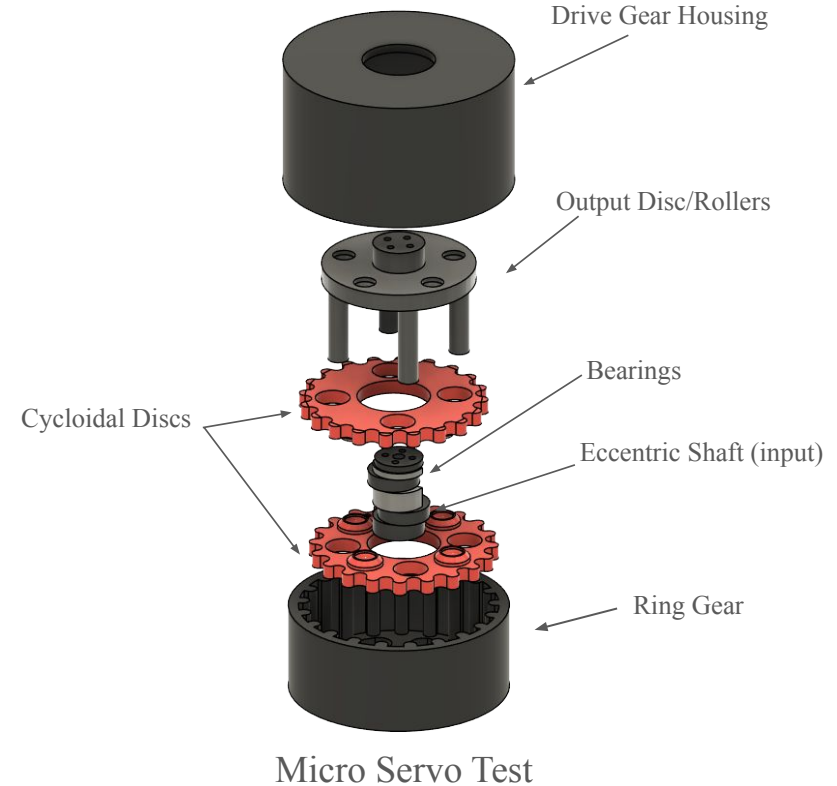


Gear Reducer

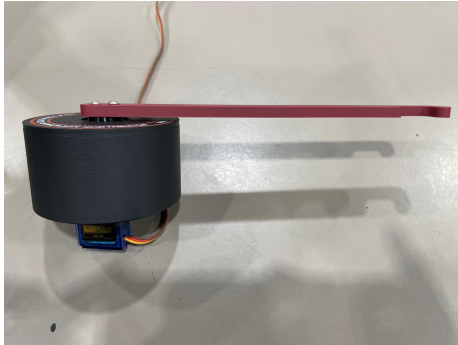
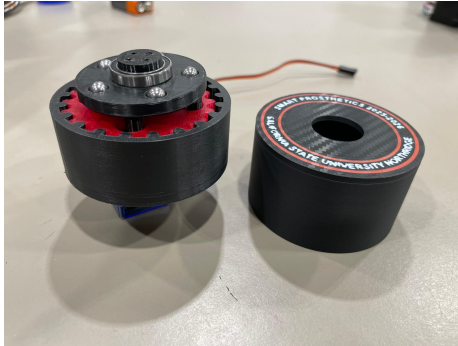
Feature	Cycloidal	Planetary	Harmonic
Reduction Ratio	High ($\leq 203:1$)	Low ($\leq 10:1$, Multi-stage for high ratio)	High ($\leq 160:1$)
Torque Density (Nm/kg)	High (≤ 76.3 Nm/kg)	Medium (≤ 44.16 Nm/kg)	Medium-High (≤ 59.14 Nm/kg)
Weight	Heavy-weight	Medium-weight	Light-weight
Backlash (Precision)	Low (≤ 1.5 arc.min)	Medium (≤ 5 arc.min)	Zero
Torsional Rigidity	High ($\leq 3,334$ Nm/arc.min)	Medium (≤ 850 Nm/arc.min)	Medium (≤ 903 Nm/arc.min)
			

Cycloidal Gear System

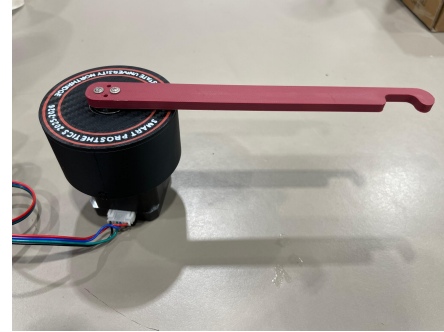
- Benefit:
 - Cycloidal drives provide high gear reduction in a smaller volume.
- Cons:
 - Servos are still limited (stepper motors typically used), gear wear, and complexity of integration.
- Micro Servo (SG90) Test
 - Transmission Ratio 21:1
 - Ideal no losses 0.15 Nm torque to 3.15 Nm
 - 70% Efficiency torque output 2.6 Nm



Cycloidal Gear System



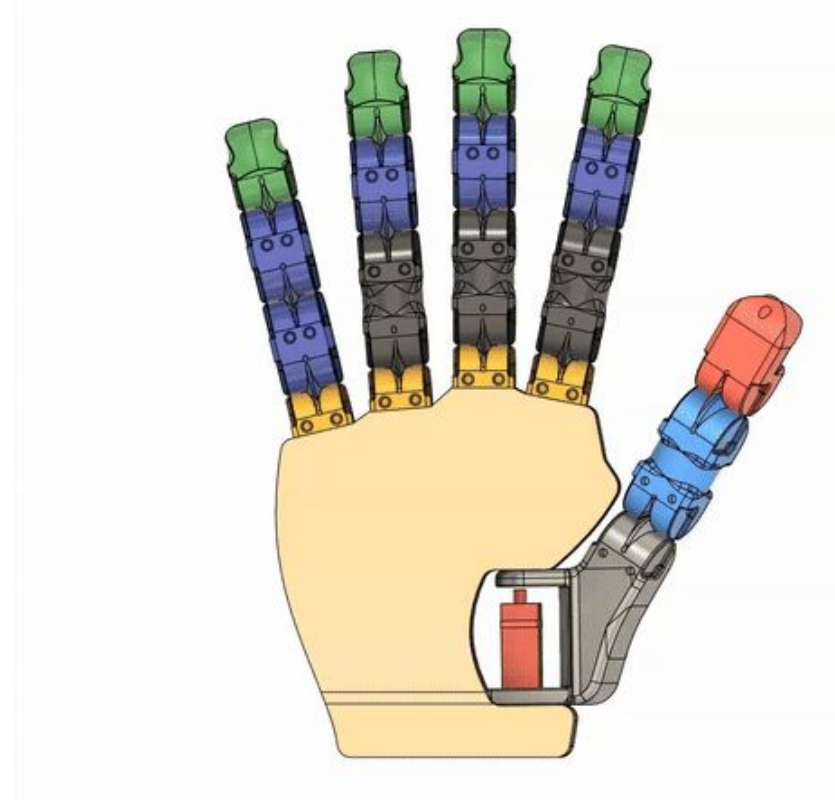
SG90 Cycloidal Gear Drive



Stepper Motor Cycloidal
Gear Drive

“Cerberus” Hand Design

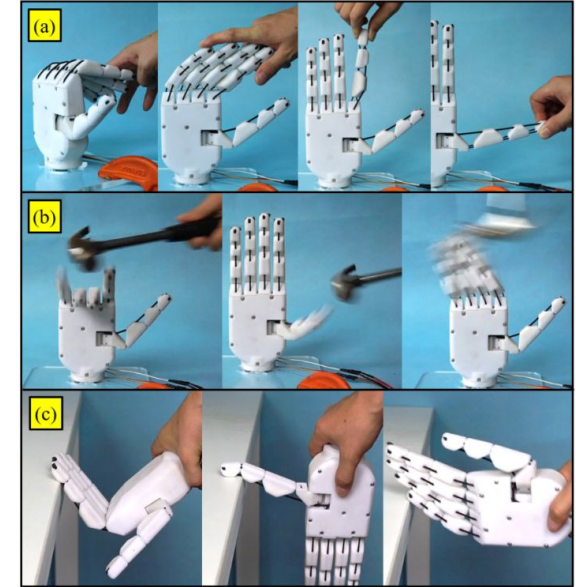
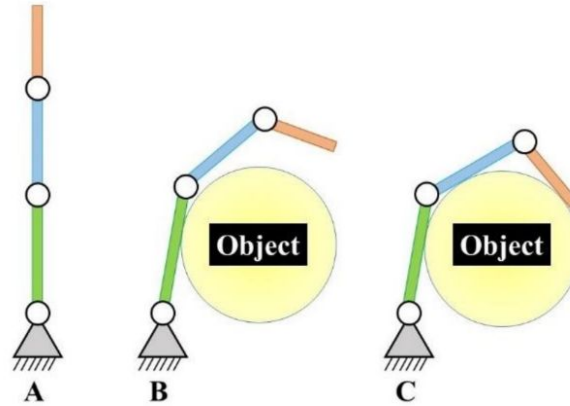
- FDM printed fingers for lighter weight
 - 45% weight reduction: 25 g $\rightarrow$ 13g
 - 8th cohort utilized SLA (Resin Printing)
- Rolling Contact Joints
- Tendon Connections
- 1 Servo / Finger
 - Lighter Weight
- 2 Servos for Thumb
 - Adduction/Abduction
 - Flexion/Extension



Joint Type

Rolling Contact Joints:

- Dislocates
- Compliant
- Biomimetic



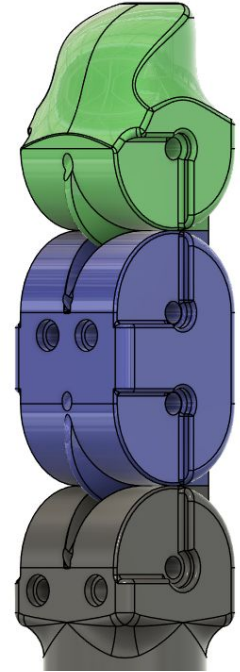
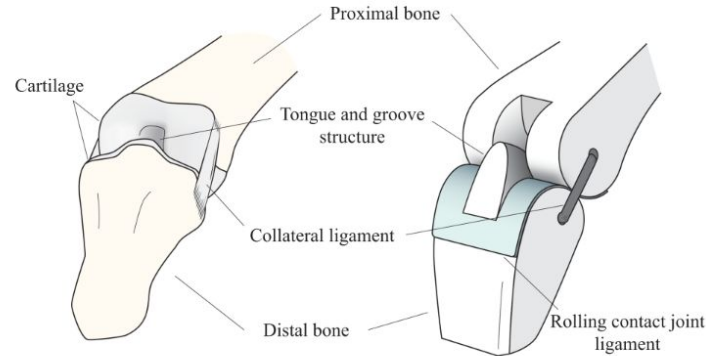
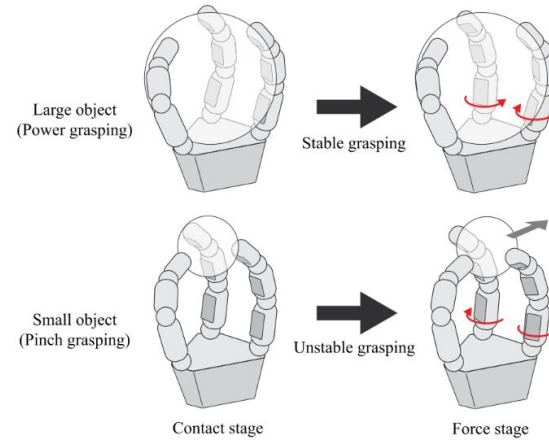
Gorjestani, Aryan & Samavati, Farzad. (2023). ATLAS Finger: a Prosthetic Finger Mechanism for Robotic applications. 9. 1-10.

H. Liu, K. Xu, B. Siciliano and F. Ficuciello, "The MERO Hand: A Mechanically Robust Anthropomorphic Prosthetic Hand using Novel Compliant Rolling Contact Joint," 2019 IEEE/ASME International Conference on Advanced Intelligent Mechatronics (AIM), Hong Kong, China, 2019, pp. 126-132, doi: 10.1109/AIM.2019.8868520.

Joint Selection Cont'd

Anthropomorphic Rolling Contact Joint:

- Tongue & Groove Design
- Torsional stiffness passively increase
- Build off 8th Cohort

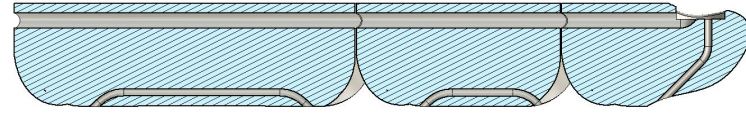


“Cerberus” Finger

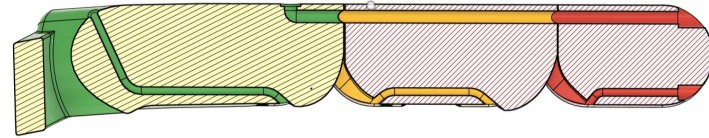
Connection Type

Tendon-Based System:

- Minimal Complexity
 - Light Weight
 - Biomimetic
-
- Flexion: Synthetic Nylon Wire
 - Extension: Elastic Band



“Cerberus” Finger



“Cerberus” Thumb



Actuation Method

Compared various
actuation mechanisms

Conclusion:

1 Motor/Finger for

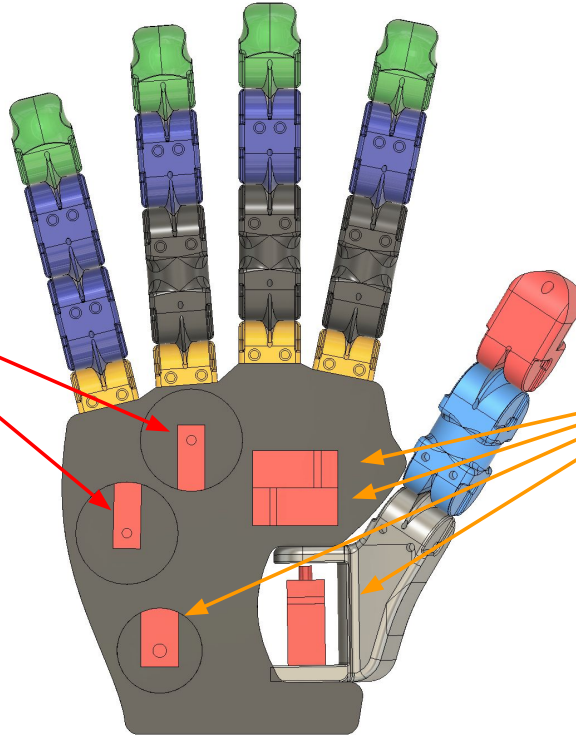
- Simplicity
- Lighter weight
- Space in palm
- Gestures

ServoMotor System					
	DOF	Pros	Cons	# of Servos	Tendon Tension/Output
1 Motor/Finger	1	Space on palm Light Weight Gestures	1 DoF	4 + (2)	6.3 kgcm ~ 10.9 kgcm
2 Motor/Finger	2	Range of Motion Grip Strength Gestures	tighter space more motors = Heavier	6 + (2)	6.3 kgcm ~ 10.9 kgcm
2 Motor/Finger (Splay)	2	Range of Motion Increase Grip Area Gestures	tighter space more motors = Heavier	6 + (2)	6.3 kgcm ~ 10.9 kgcm
MERO Hand	1	2 motors for whole hand Simple mechanism	Strength Limited Gestures(Close/Open)	1 + (1)	3.6 N (avg grasp)
Differential Pulley	1	Even distribution of load with 1 motor Adapting to item	1 DoF Thumb Design Limited gestures: Close/Open Takes Space Friction in tendon and pulley	1 + (1)	24.19 +/- 1.91 N
Four Output Gear Differential	1	Even distributon of load Requires 2 motors	Complicated Design Not compact Heavy (169.8g) Limited Gestures (x4)	2 + (2)	39 N
Whiffletree mechanism	1	Even distribution of load Total of 16 finger combinations	Manually lock differential mechanism, Require additional space precision tension calibration	1 + (1)	42.8 N

Servo Motor Selection



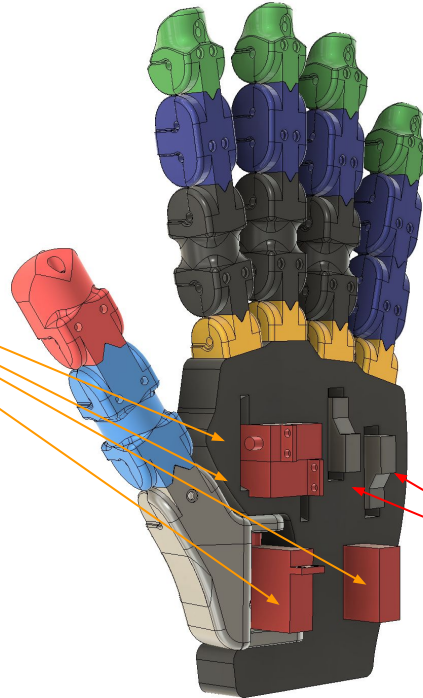
KST A08



MKS HV6150H

Servo Motor Selection

	Torque (kgf.cm)	Voltage (V)	Weight (g)	Length (mm)	Width (mm)	Height (mm)
KST A08	1.4	3.8	7	23.5	8	19
	2.2	6				
	2.8	7.4				
	3.2	8.4				
MKS HV6150H	6.3	4.8	17.4	23.5	11	25
	8.2	6				
	9.8	7.4				
	10.9	8.2				



4x MKS motors: 69.6 g

2x KST motors: 14 g

Total weight of Servos: **83.6 g**

Servo Motor Selection

Name	Torque (kgf.cm)	Voltage (V)	Weight (g)	Length (mm)	Width (mm)	Height (mm)
KST A08 LV	1.2	3.5	7	23.5	8	19
	2.0	4.2				
KST HS08B HV	2.8	3.8	11	23.5	8	25.25
	4.5	6				
	5.2	7.4				
	6	8.4				
KST X08	1.4	3.9	8	23.5	8	16.8
	2.2	6				
	2.8	8.4				
KST X08 Plus	2.4	3.8	9	23.5	8	16.8
	3.85	6				
	4.85	7.4				
	5.3	8.4				
KST X10	6.2	4.8	28	30	10	35.5
	7.5	6				
	9.5	7.4				
	10.8	8.4				

Future Works

- Attachment:
 - Develop/ build the upper arm assembly
 - Develop the electronics for the Adaptive Attachment System
- Forearm:
 - Testing Nylon 6 carbon fiber
 - Build forearm assembly
- Hand:
 - Manufacture and test the grip strength of the hand
 - Collaborate with Foot Control to assign various unique gestures

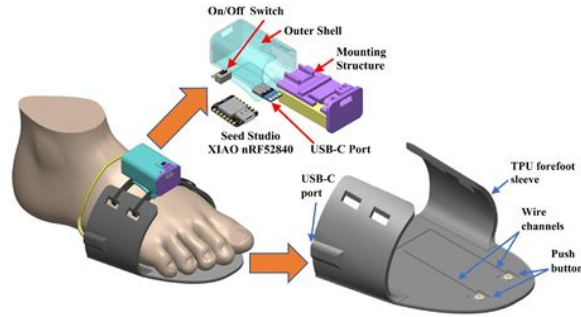
Foot Controller System

Team Members:
Kurt Trocino
Kainen Shaw
Matthew Portillo
Hao Phan

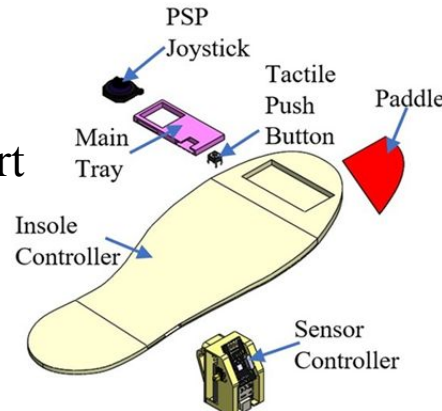
Smart Prosthetic Foot Controller History

- Noninvasive
- Low Cost
- Ease of use

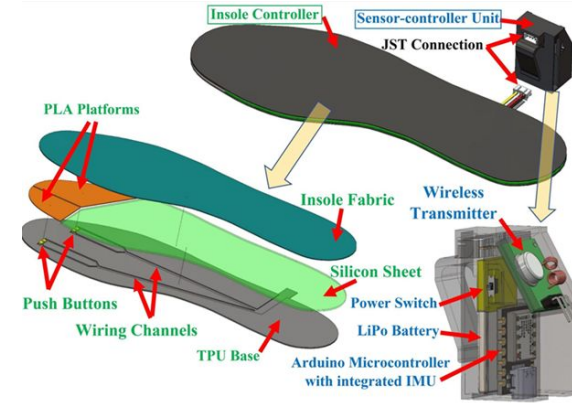
7th Cohort



6th Cohort



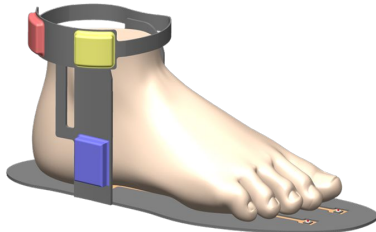
5th Cohort



8th Cohort Foot Controller Pros and Cons

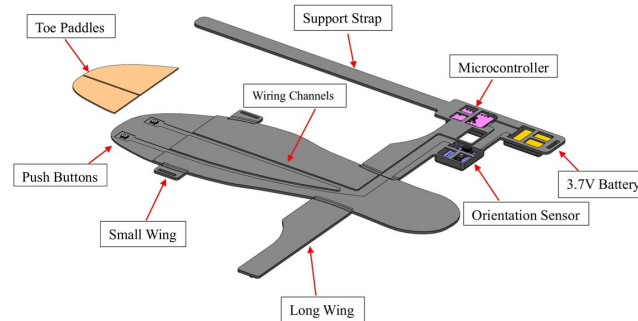
Pros:

- Thin, easy to put into shoe
- Does not affect walking
- Relatively easy to use



Cons:

- The IMU fixture is not rigid
- The strap does not securely affix to user's body
- Shoe dependent



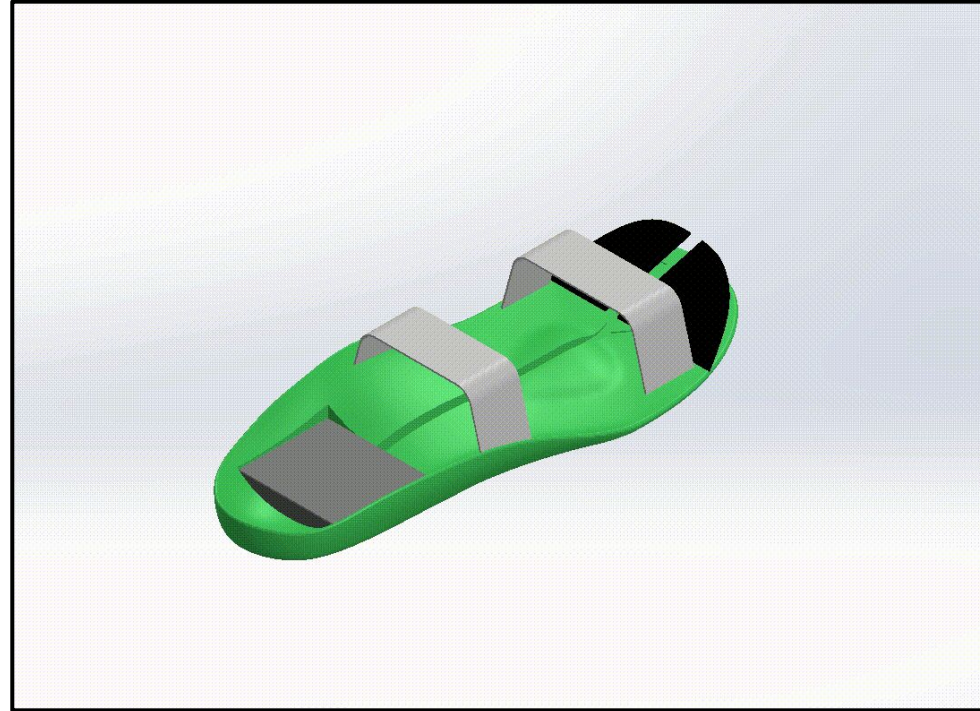
9th Cohort Design Goals and Objectives

- Redesign IMU fixture to improve output quality
- Improve ergonomics of foot controller layout
- Optimize neural network for gesture recognition
- Increase control method options of prosthetic



Foot Controller Redesign

- Relocated IMU for improved data
- Case: PC for high impact resistance, strength, durability, and thermal resistance
- Improved ergonomics
- Resized toe button flappers and relocated small toes button
- Elastic straps to make foot controller wearable barefoot



Foot Controller Redesign

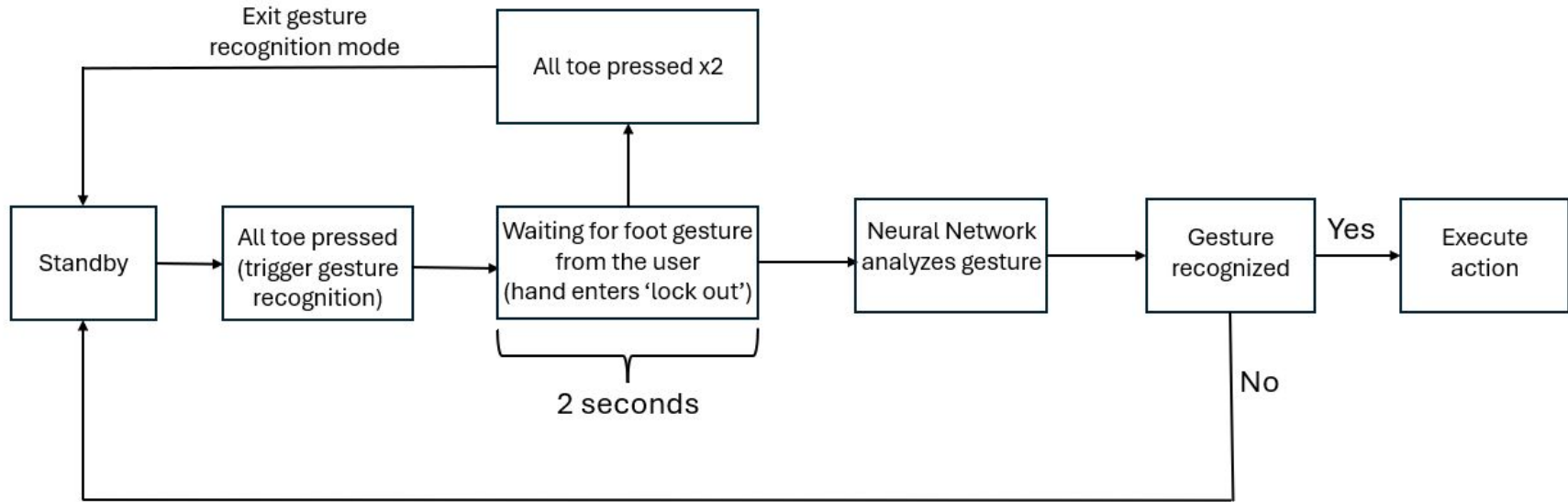
- Pros:
 - Improved IMU data
 - Robustness of insole
 - Intentional strengthening of battery housing and easier removal
- Cons:
 - Thickness of heel



Buttons Function

Big toe press and hold 	Small toe press and hold 	All toe press 	Big toe press → small toe press 	Small toe press → big toe press 
Close hand gradually 	Open hand gradually 	Trigger gesture recognition mode	Cycle forward grip  	Cycle backward grip  

Gesture Recognition Mode

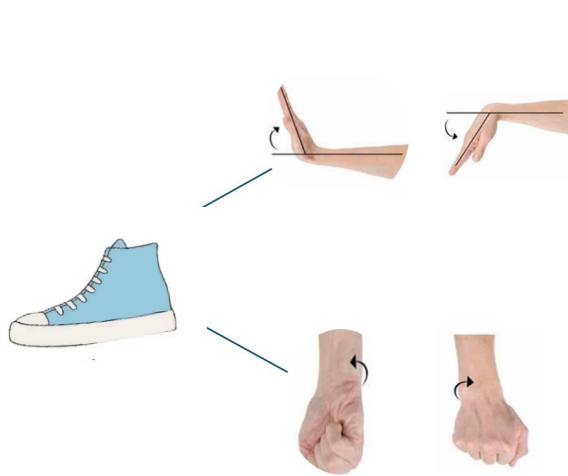


List of Gestures

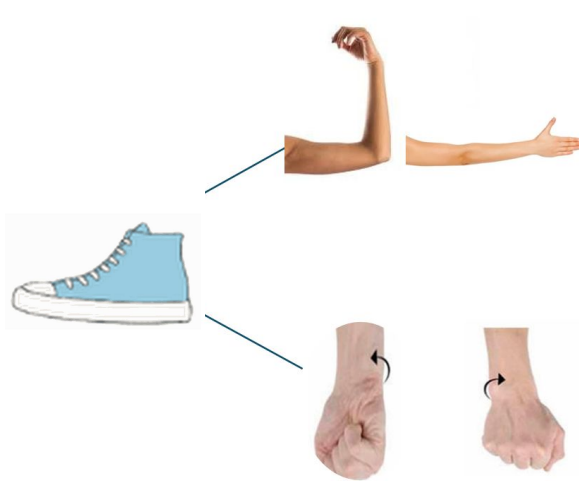
Toe tap		Wrist Actuation Mode
Flat tap		Elbow Actuation Mode
Heel tap		Shoulder Actuation Mode
Swipe left		Decrease servo motor speed
Swipe right		Increase servo motor speed
Heel click		Return to default

*Animations made from
real IMU readings*

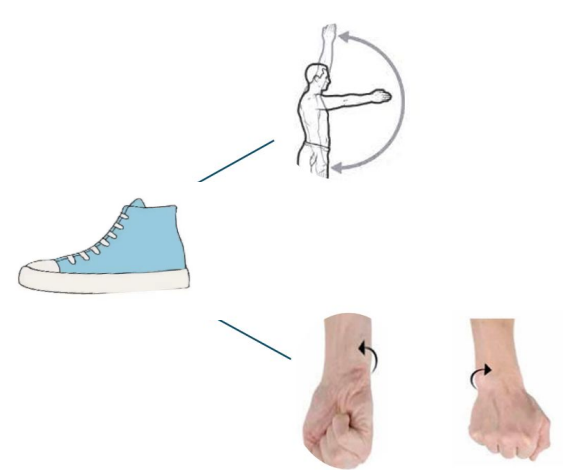
Actuation Modes



Wrist actuation mode

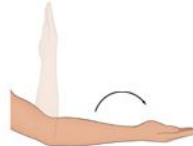


Elbow actuation mode

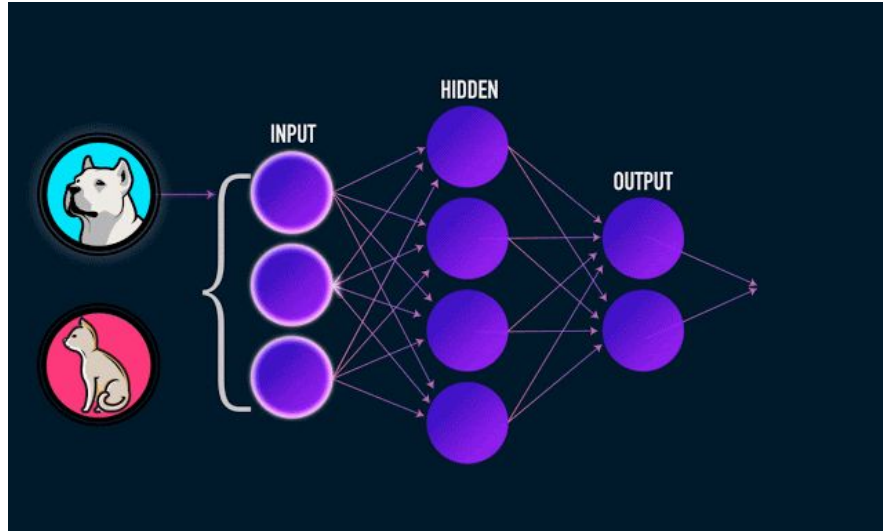


Shoulder actuation mode

Controlling the Arm Movement

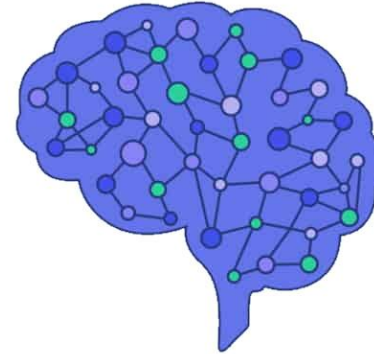
Foot motion	Ankle dorsiflexion 	Ankle plantarflexion 	Ankle inversion 	Ankle eversion 
Wrist actuation mode	Wrist extension 	Wrist flexion 	Wrist supination 	Wrist pronation 
Elbow actuation mode	Elbow flexion 	Elbow extension 	Wrist supination 	Wrist pronation 
Shoulder actuation mode	Shoulder flexion 	Shoulder extension 	Wrist supination 	Wrist pronation 

What is a Neural Network?

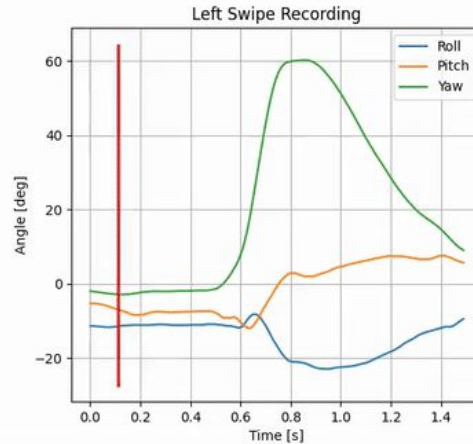
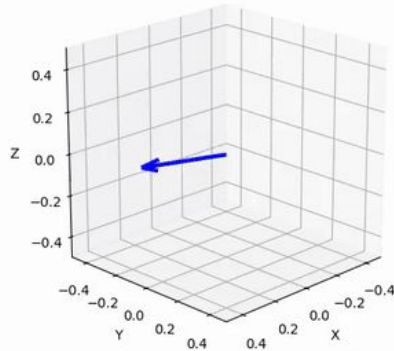


Neural Network Binary Classification
between Dog and Cat

A Neural Network is an algorithm that can be “trained” to recognize and classify patterns from data.



Neural Network for Gesture Recognition



Example of a “Left Swipe” Gesture IMU recording

Goal: Train a neural network to distinguish between all six gestures reliably in real time using IMU data as input.

Convolutional Neural Network (CNN)

Compared to a Feed Forward Neural Network:

Pros:

Requires Less Training Data

Accounts for Sequential or Time-Based Patterns

Multi-channel Input

Cons:

Higher Complexity in Setup

Computation Cost (TBD)

1 x1	1 x0	1 x1	0	0
0 x0	1 x1	1 x0	1	0
0 x1	0 x0	1 x1	1	1
0	0	1	1	0
0	1	1	0	0

Image

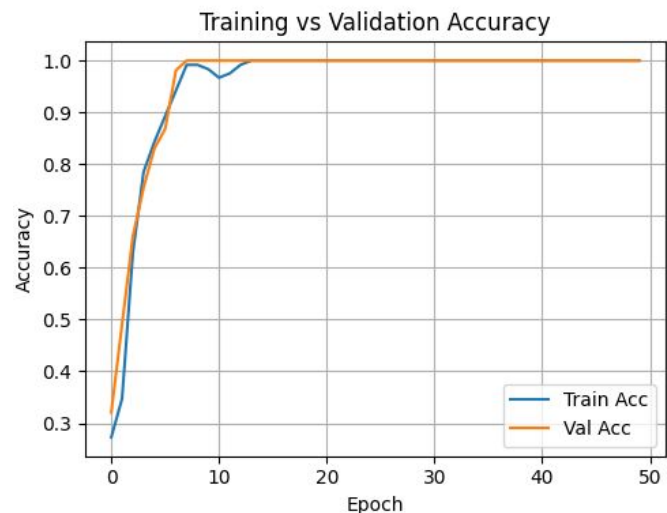
4		

Convolved
Feature

Convolution Operation Visualization

Preliminary Model Training

- As a test, we recorded a dataset containing only four distinct gestures
- After some tuning, the CNN model labeled all 174 samples with 100% accuracy.
- Average inference time is 58.44ms on a computer.



Accuracy of Neural Network during Training

Tasks Before CDR

- Have a working prototype of the redesigned foot controller
- Test the Convolutional Neural Network model in real time
- Reduce redundancies and improve efficiency of arm and foot controller code

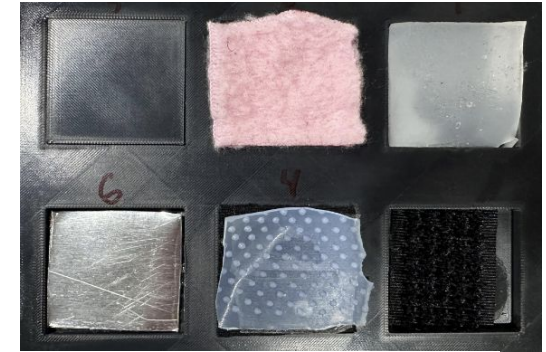
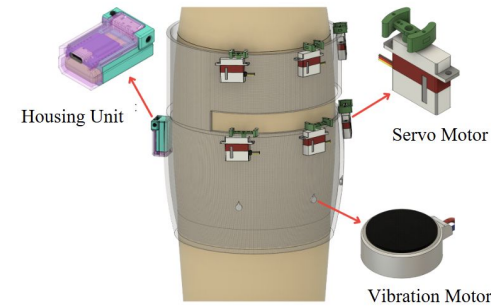
Haptic Feedback

Team Members:

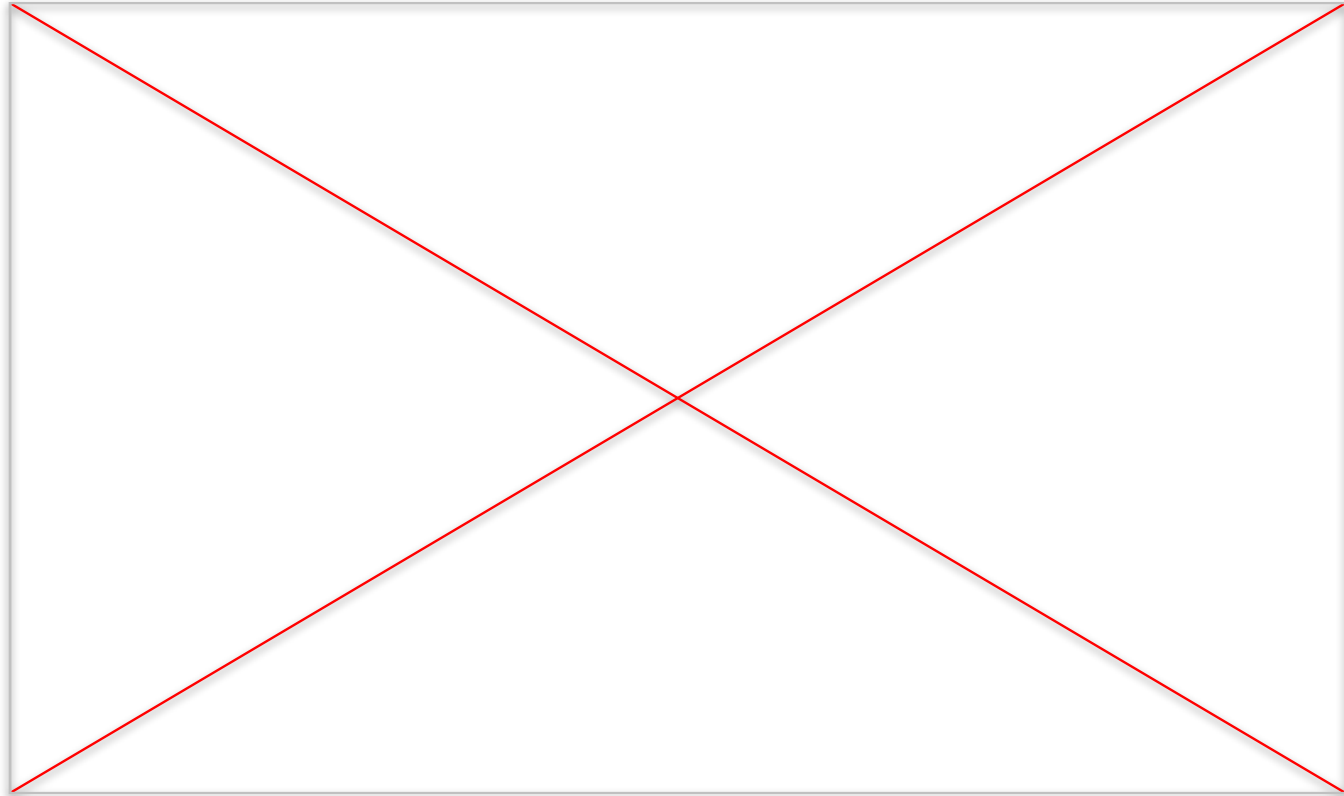
Jorge Alvarez, Nathan Handal, Chris Lopez, Vanessa Montejano

8th Cohort

- Motivation: Create noninvasive design for amputees who experience lack of sensory touch and can also serve as an addition to other prostheses in the market
- Design Overview
 - Used multiple servos to control texture application
 - Used a camera to detect different materials and applied closest material on upper arm
 - Electret microphone housed in a fingerprint-inspired case to pick up frequencies

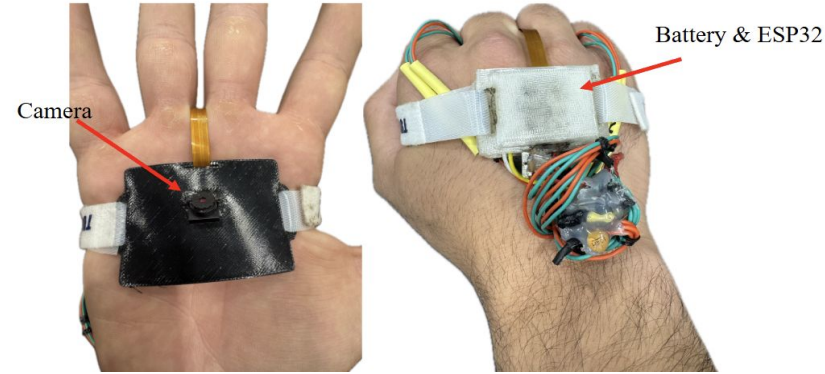


8th Cohort



8th Cohort

- Limitations
 - Increased size
 - Wiring complexity
- Problems
 - Visual assistance of the materials was necessary for user to understand the material being applied
 - Unable to get camera to work so both systems weren't able to connect



Goals and Improvements

- Goal
 - Develop compact sleeve capable of providing texture application with fewer components
- Improvements
 - Camera successfully works
 - Exoskeleton frame attaches directly to wearable robotic arm.
 - IR temperature sensors on fingertips for heat-based feedback.
 - Compact conveyor texture system replaces servo-based application.
 - BOA/Dial-based tightening system for quick, secure fit.
 - Fully detachable sleeve modules for maintenance and upgrades.
 - Enhanced vibration and thermal feedback synchronized with visual input.

8th Cohort vs Current Design Comparison		
Category	8th Cohort Design (2024–2025)	Current Design (Your Team)
System Overview	Two subsystems: Texture Identification (EchoPrint + camera) and Texture Application.	Integrated multi-sensor haptic feedback system with improved texture detection and real-time
Sensors & Microcontroller	SEEED XIAO ESP32-S3 with accelerometers and microphone; no temperature sensing.	SEEED XIAO ESP32-S3, accelerometers and IR temperature sensors on fingertips.
Texture Identification	EchoPrint system with electret microphone and camera for roughness and material detection.	Dual-sensor setup (optical + temperature) with accelerometer input for enhanced tactile and thermal mapping.
Texture Application	Servo motor-based armband applying interchangeable textures (Velcro, silicone, fabric, rubber, metal).	Compact conveyor belt system on the sleeve for texture delivery; modular and detachable
Wearable Design	Large silicone finger sacks, Velcro straps; non-detachable assembly.	Lightweight wearable sleeve with BOA tightening system for quick wear/removal; separate clip-on
Feedback Mechanism	Vibration and material contact on upper arm.	Vibration feedback plus temperature feedback for realistic touch
Form Factor	Bulky glove-style with embedded electronics and wiring.	Compact modular sleeve design with detachable components and integrated cable routing.
Software & AI Model	Edge Impulse grayscale model (~85% accuracy) for material classification.	TensorFlow Lite / Edge Impulse hybrid model with adaptive real-time learning.
Testing Results	57–69% tactile accuracy post-training; 84.9% visual classification accuracy.	In Progress

M.I.D.A.S. System Architecture

Hardware Development	AI Integration	System Testing & Validation
Glove and sleeve prototypes under fabrication and	CNN model training → TensorFlow Lite deployment	Thermal, vibration, texture trials to verify perception and

Project Objective

**Realistic
Texture
Interaction**

**Integrated
Compact
Design**

**Enhanced
Sensory
Feedback**

**Improved
Modularity
and User
Experience**

Core Hardware Components

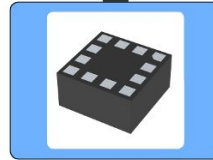
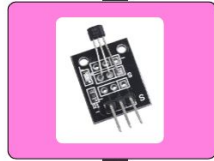


Diagram key

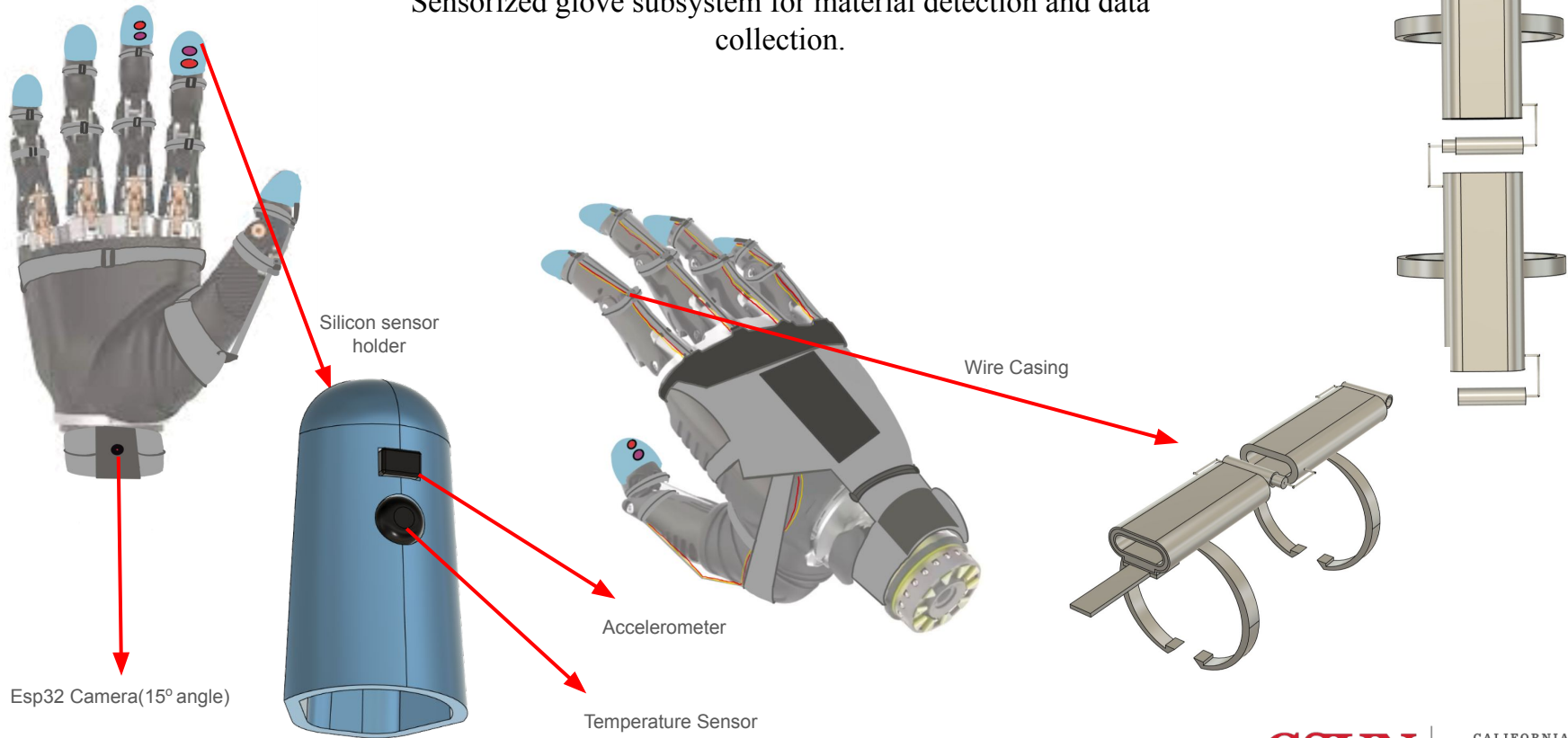
- Both Systems
- Sleeve
- Glove



M.I.D.A.S. Gauntlet Subsystem

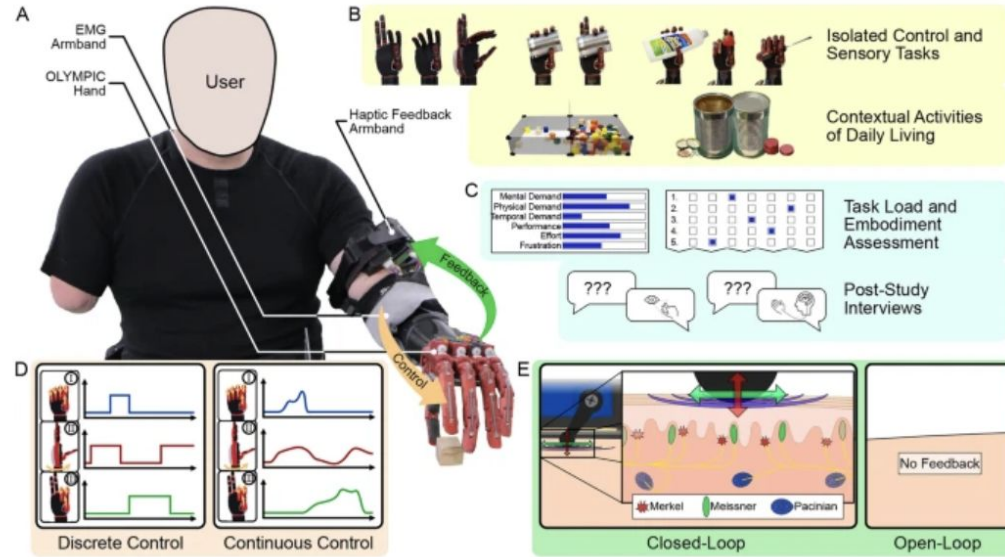
(Multimodal Integrated Detection and Stimulation)

Sensorized glove subsystem for material detection and data collection.



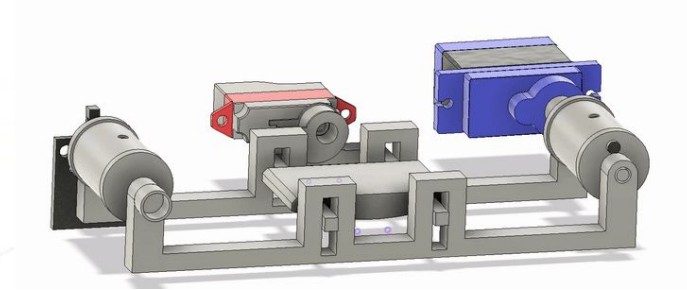
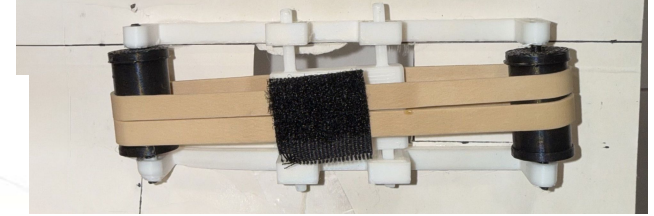
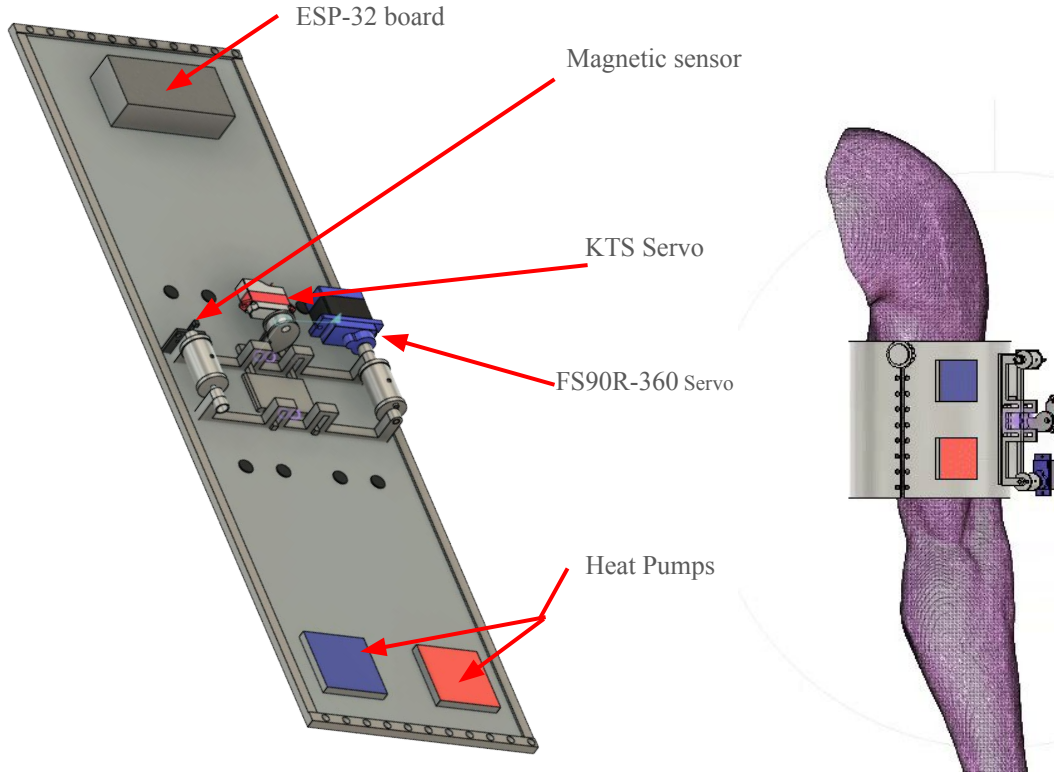
Haptic Feedback Armband

- Closed-loop control improves performance.
- Multimodal feedback enhances realism.
- Supports our system design.



M.I.D.A.S Conduit Subsystem

Texture-application and multimodal feedback sleeve for upper-arm testing.



Electronics Architecture

Component	Subsystem	Function	Voltage / Current Specs
Seed XIAO ESP32-S3 Sense	Both	Main controller, Wi-Fi, ML inference	5 V in / 3.3 V logic / < 250 mA
MXC6655XA (Vibrational sensor)	Glove	Motion Sensing	1.62 - 3.6 V / < 0.3 mA
Thermal Camera Module	Glove	Thermal Detection	5 V / ~150 mA
TEC1-12706 (T.E. H/C Pump)	Sleeve	Heating / cooling actuator	12 V / 6 A max
Vibration Motors	Sleeve	Vibrotactile feedback	3 V / 75 mA each
FEETECH Servo	Sleeve	Texture applicator motion	5 V / 0.8 A stall
Magnetic Sensor	Sleeve	Position Feedback	5 V / < 20 mA
Li-Po Battery	Both	Power supply (3.7 V / 2000 mAh)	Boosted to 12 V and 5 V rails

- Power budget assumes 85 % from the DC-DC step-up regulators supplying 12 V and 5 V rails.
- TEC1-12706 current limited to ≤ 1.5 A to maintain safe operation and realistic runtime.
- Both glove and sleeve powered by independent 3.7 V / 2000 mAh Li-Po packs.

$$P_{\text{Sleeve}} = (12V)(1.5A) + (5V)(0.8A) + (3V)(0.075A \times 4)$$

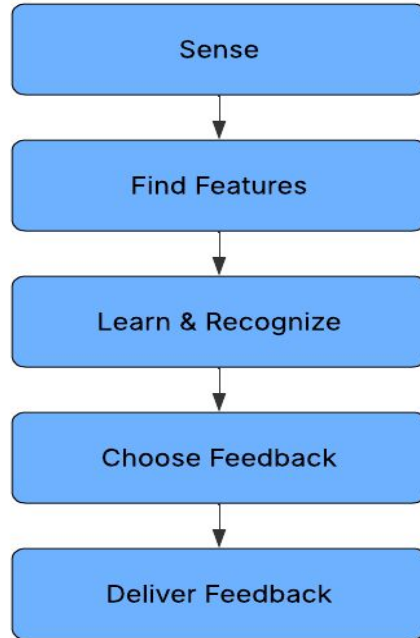
$$P_{\text{Sleeve}} = 21W$$

$$I_{\text{battery}} = \frac{P_{\text{Sleeve}}}{\eta \cdot V_{\text{battery}}} = \frac{21}{0.85 \cdot 3.7} = 6.52A$$

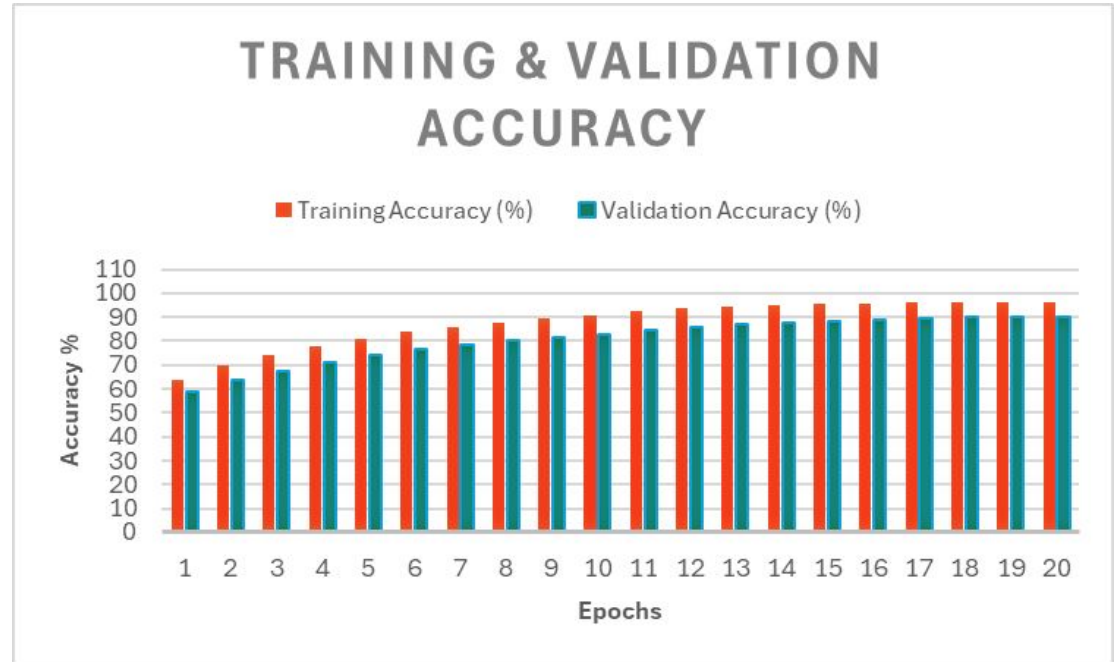
$$t_{\text{runtime}} = \frac{2Ah}{6.52A} = 0.31h$$

$$t_{\text{runtime}} = 0.31h \cdot \frac{60mins}{1h} = 18.6min$$

TensorFlow Implementation



Simulates how accuracy improves as the model learns.



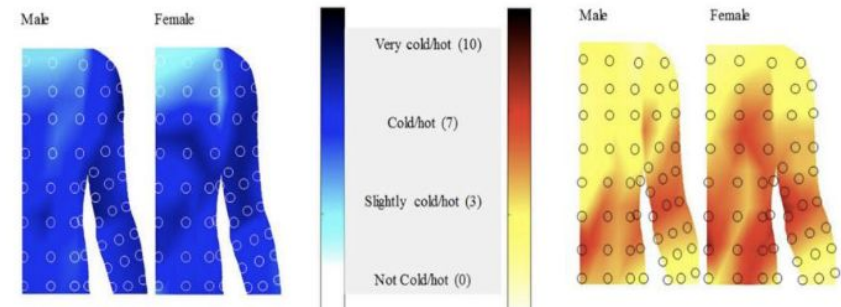
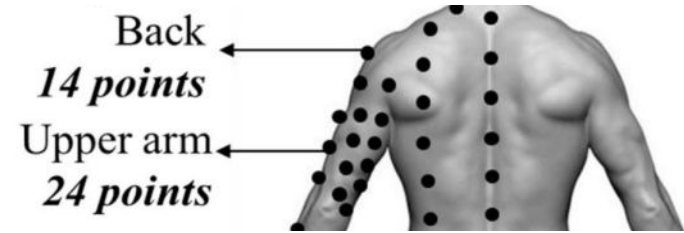
```
epochs = np.arange(1, 31)
train_acc = 0.6 + 0.4*(1 - np.exp(-0.15*epochs))
val_acc = 0.55 + 0.38*(1 - np.exp(-0.12*epochs))
plt.plot(epochs, train_acc*100, label='Train')
plt.plot(epochs, val_acc*100, label='Validation')
```

Experimental Validation Plan

Purpose: Train participants to distinguish textures through touch and vision to simulate haptic feedback for the prosthetic arm

How: Controlled stimuli (temperature, vibration, and texture) applied to mapped skin regions using the Conduit Sleeve system. Participant responses recorded to evaluate perception accuracy and comfort

Why: Understanding how humans integrate touch and vision to distinguish textures supports the design of prosthetic systems that provide natural, intuitive haptic feedback for improved control and embodiment



Future Iterations & Path Forward

- Will work on connecting vibration sensors as well as the thermal sensors for the sleeve
- Create prototype for the texture application conveyer belt with servos and textures
- Begin training to gather data on best spot for texture identification
- Improving battery life
- Figuring out cold temperature application

9th Cohort Calendar

- Team Invites: June 6th, 2025
- Research Papers: August 30th, 2025
- NBC Interview: September 2nd, 2025
- PDR: October 22nd, 2025
- CDR: November 26th, 2025
- Cal State Biotech Fair: January 24th, 2026
- Report: March 22nd, 2026
- CSUN Posium: April 2026
- CSUN Showcase: May 2025

Thank you!
Any Questions?