

# **Adaptive Hip Exoskeleton Control for Personalized Assistance Using Deep Learning**

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# Research Vision

Enable wearable robots to provide personalized, tuning-free assistance by estimating human physiological states in real-time using deep learning

## Key Contributions

**AI-Driven Control**  
Temporal Convolutional Network (TCN) estimates biological joint moments from wearable IMUs in real-time

**Clinical Validation**  
First tuning-free hip exoskeleton tested across diverse daily activities in stroke survivors (N=13)

**Hardware Design**  
Designed and fabricated hip exoskeleton optimized for clinical populations

**Open-Source Dataset**  
Comprehensive stroke biomechanics dataset (N=20, 30 activities, multi-modal sensors) to be released publicly

# The Problem: Stroke & Mobility

## Global Burden

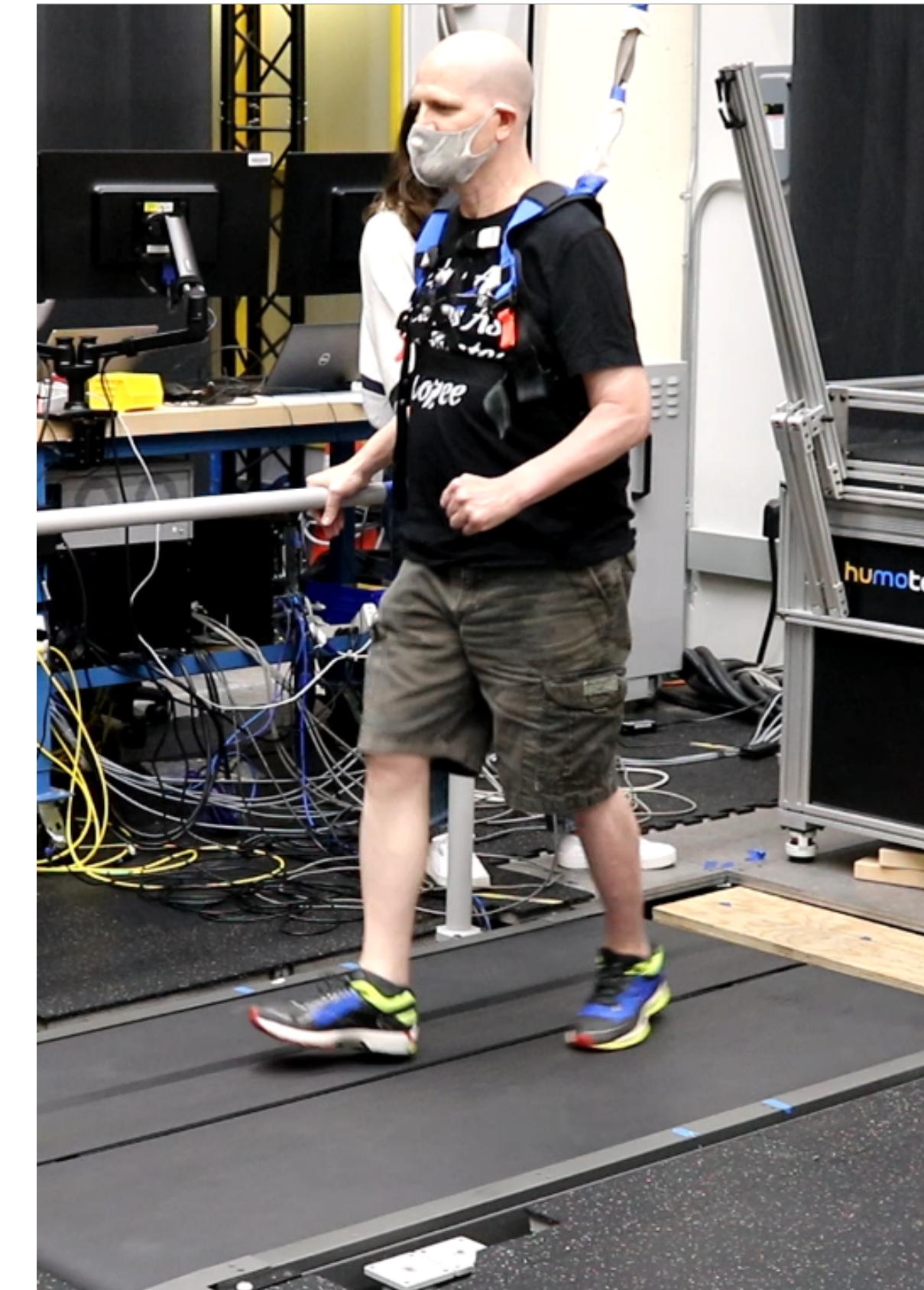
- 1 in 4 people will experience stroke
- 94M survivors
- #2 cause of death and disability

## Impact on mobility

- 50-70% reduction in walking speed
- 80% gait asymmetry
- 1.5-2x energy cost
- 2x fall risk

## The Gap

7M chronic stroke survivors in US have limited rehab access after insurance ends. Assistive robots could restore independence



“Assistive wearables could help—but require extensive manual tuning”

# Technical Challenge: Why Personalization is Hard

## Current Exoskeleton Limitations

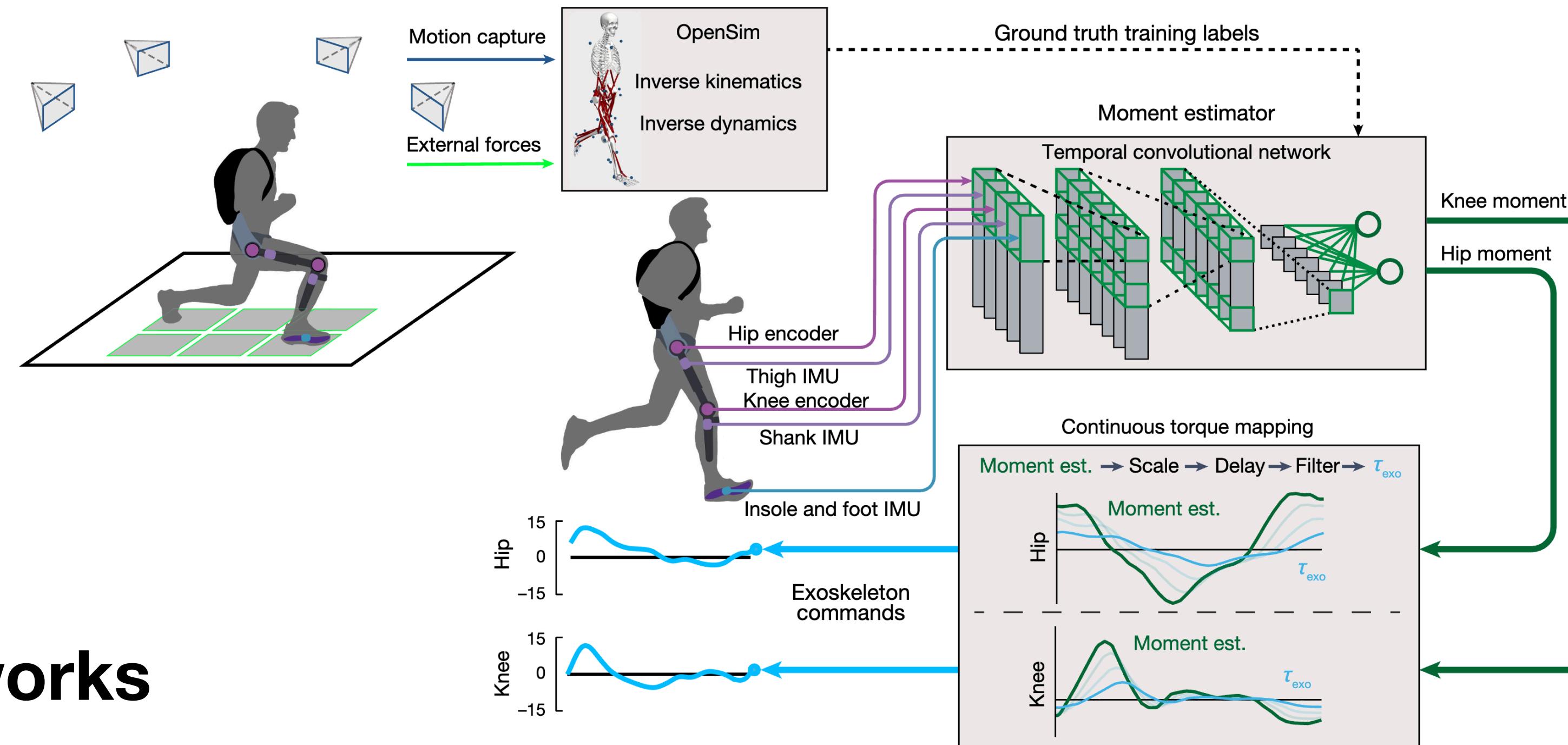
Challenge	Current Approach	Problem
Tuning Adaptation Activities Outcomes	Manual parameter adjustment Fixed assistance profiles Limited tasks Highly variable	Required expert; time-consuming Can't adapt to user variability Limited real-world utility Some users don't benefit

**Research Question:** “How can we provide personalized assistance that adapts to each user’s unique needs without manual tuning?”

**My Solution:** Use deep learning to continuously estimate the user’s biological joint moments from wearable sensors, then provide proportional assistance in real-time

# My Approach: Biological Moment-Based Control

**Concept:** Instead of pre-programmed assistance patterns, estimate what the user's muscles are trying to do, then assist accordingly



## Why this works

### User-independent

Single model works across different users without retraining

### Task-agnostic

Seamlessly adapts to walk, stairs, sit-to-stand, and other activities

### Volitional

Assistance follows user intent continuously

# Method: TCN for Joint Moment Estimation

## Input

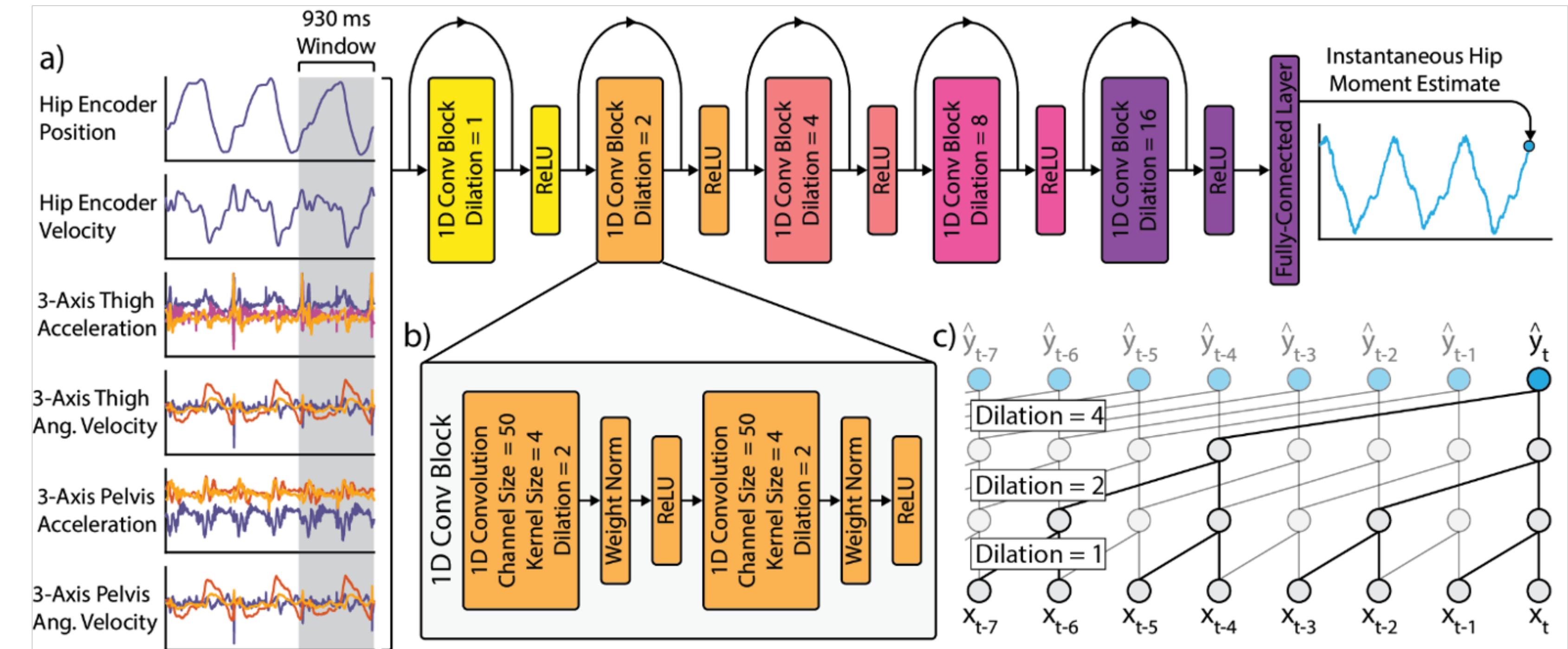
- 7 IMUs (pelvis, thighs, shanks, foot)
- 2 insoles
- 44 features

## Architecture

- Temporal Convolutional Network (TCN)
- Causal convolutions
- ~5ms inference

## Output

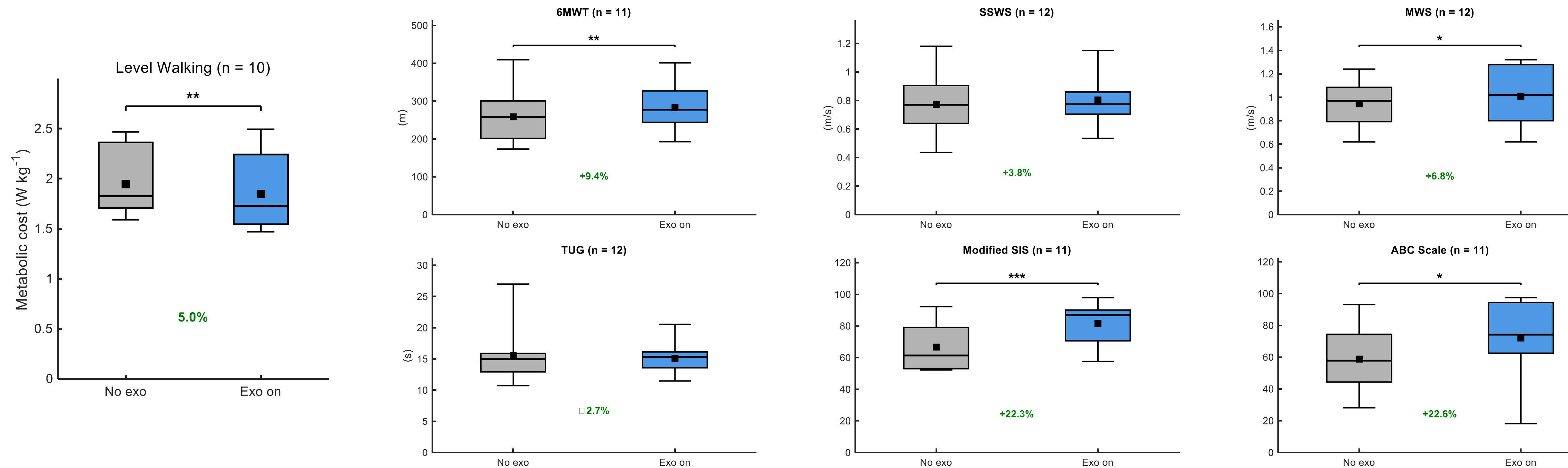
- Bilateral hip moments (Nm/kg)



**Training Data:** N=20 stoke, 39 activities, motion capture + force plates + IMUs

# Results: Clinical Trial (N=13 Chronic Stroke)

**Study Design: 2-day protocol, No Exo vs. Exo with TCN-base control**



**Improved metabolic costs, gait quality, and quality of life**

**Key Insight:** The same controller, without any tuning, automatically adapted to individual deficits without any tuning

# Hardware: Hip Exoskeleton Design

## Specifications

- Bilateral hip flex/extension assistance
- Peak torque: 15 Nm
- Weight: 2.68 kg

## Clinical Features

- Adjustable hip width for diverse body types
- Quick-adjust thigh cuffs for easy donning
- Passive ab/adduction to not restrict natural movement
- “Second Skin” sensor integration for comfortable IMU placement

I designed and fabricated this system, iterating based on feedback from stroke survivors and clinicians



# Dataset Contribution: Open-Source Resource

## Dataset Overview

## Participants

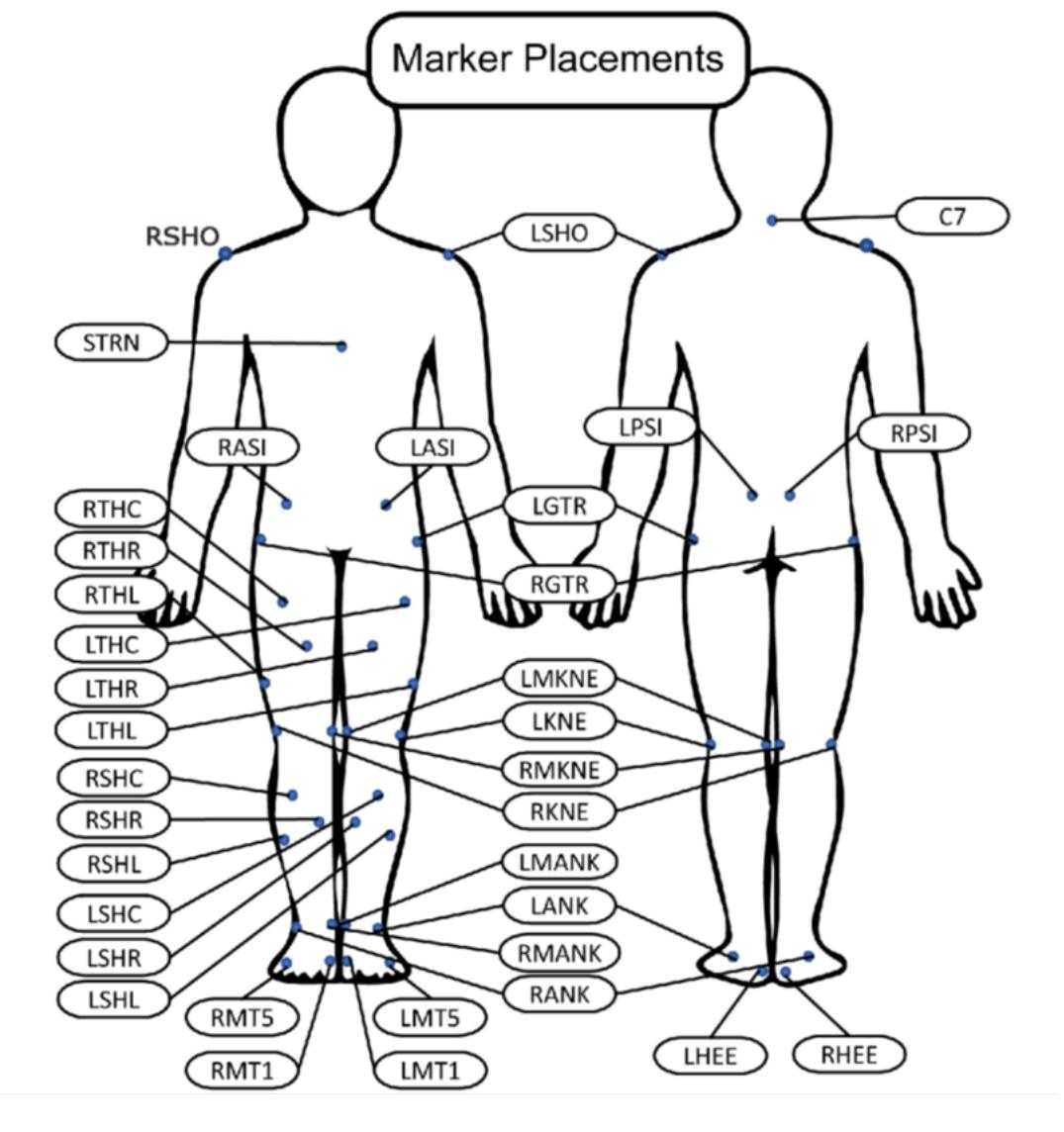
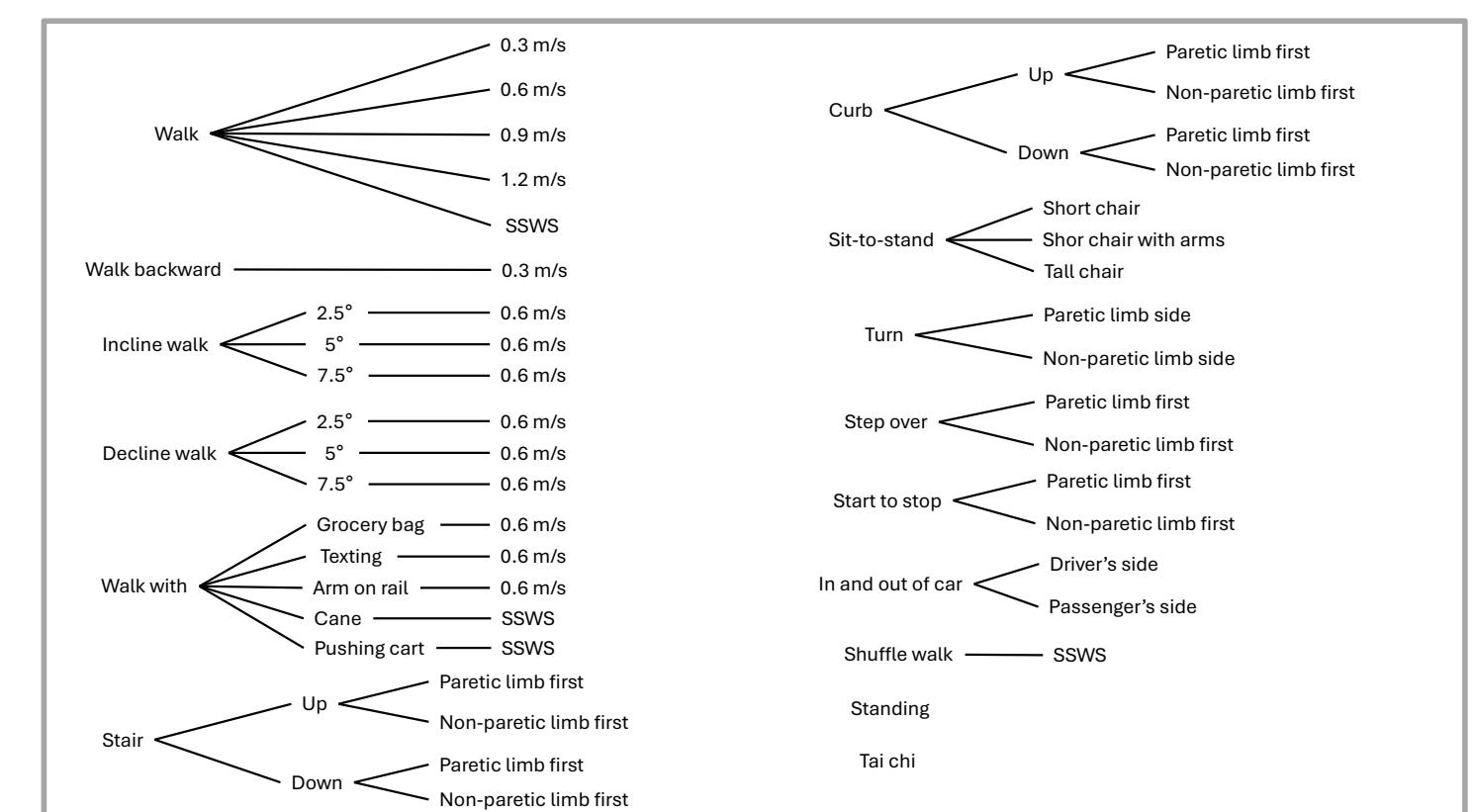
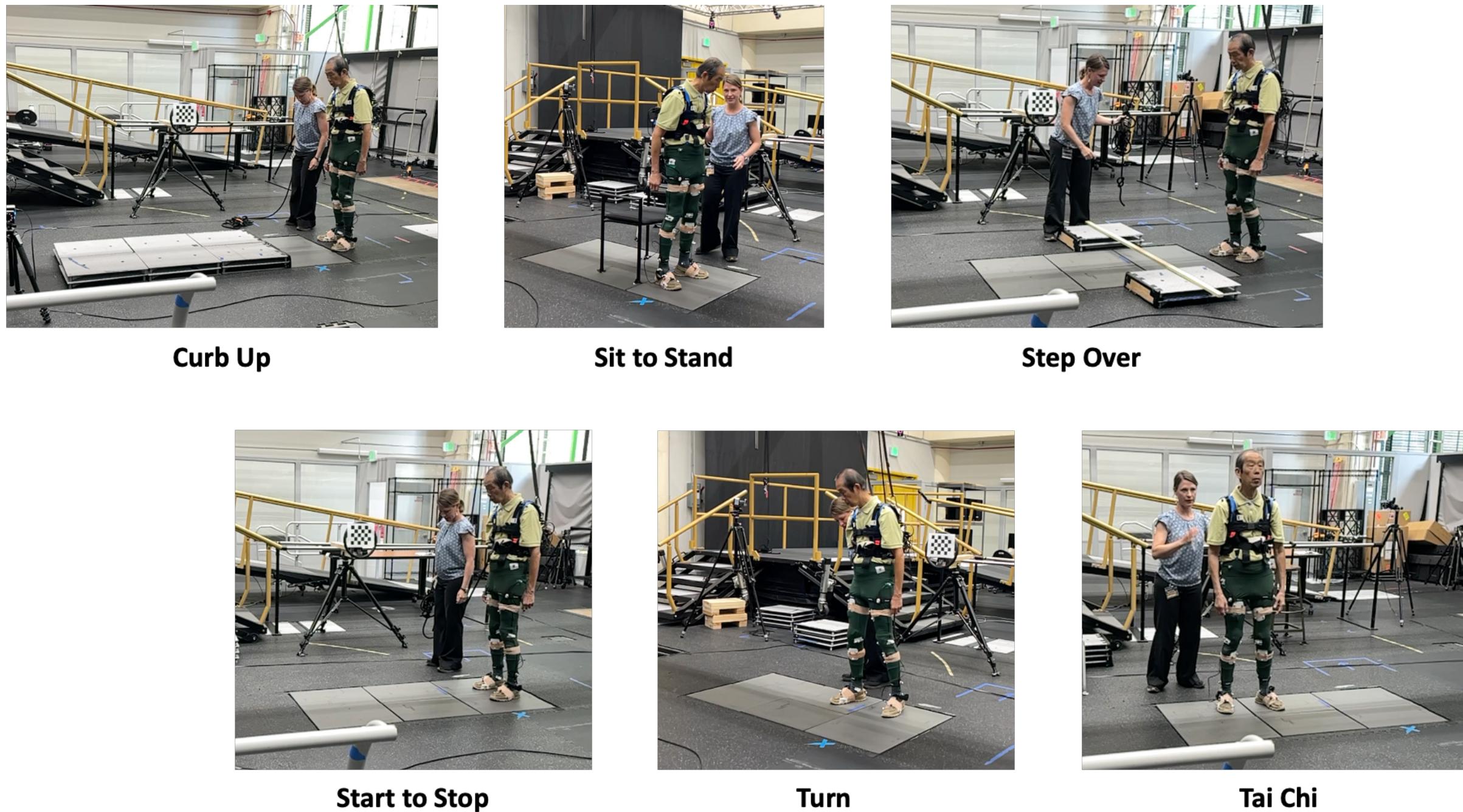
- N=20 stroke + N=5 able-bodied controls

## Activities

- 39 daily living tasks

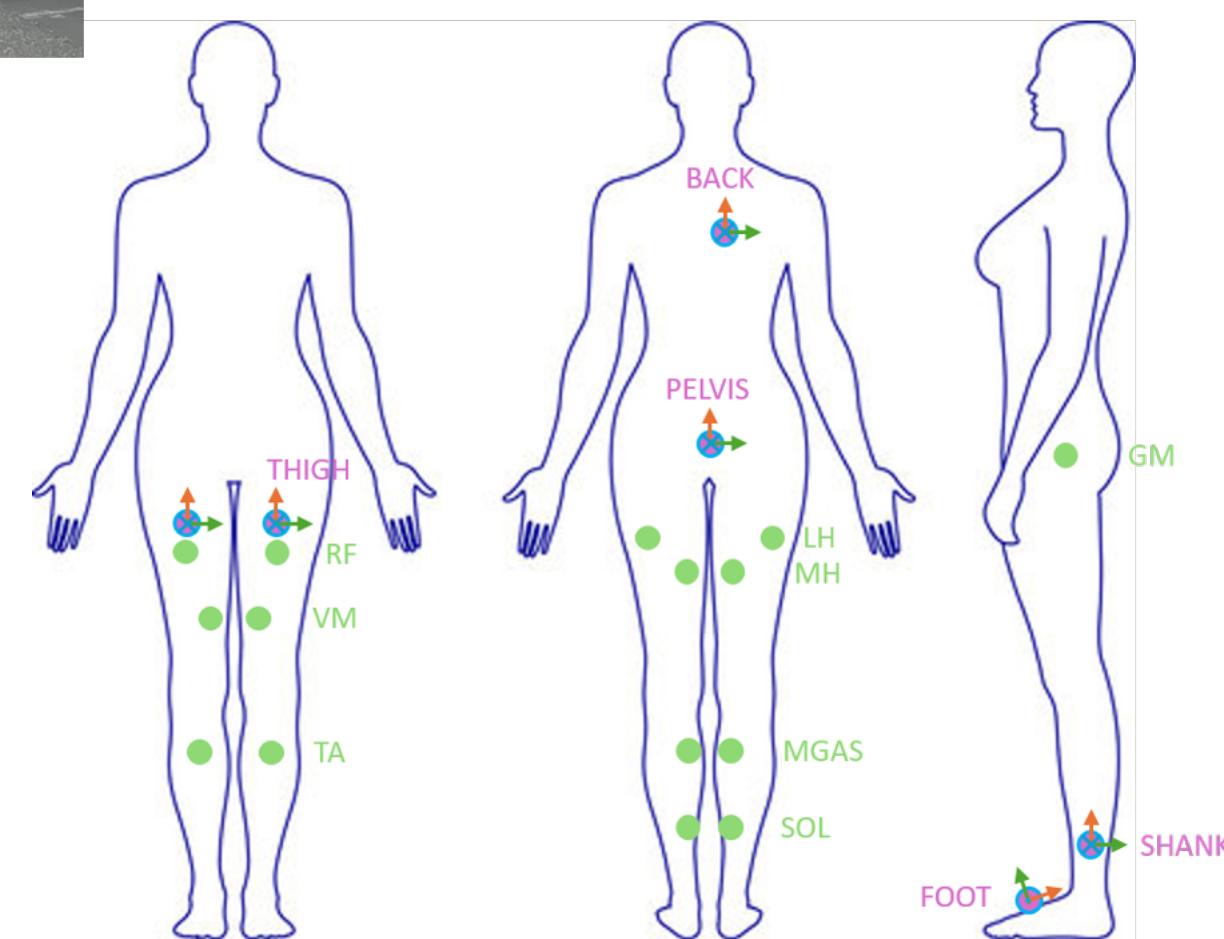
## Modalities

- Motion capture
- Force plates
- IMUs
- EMG

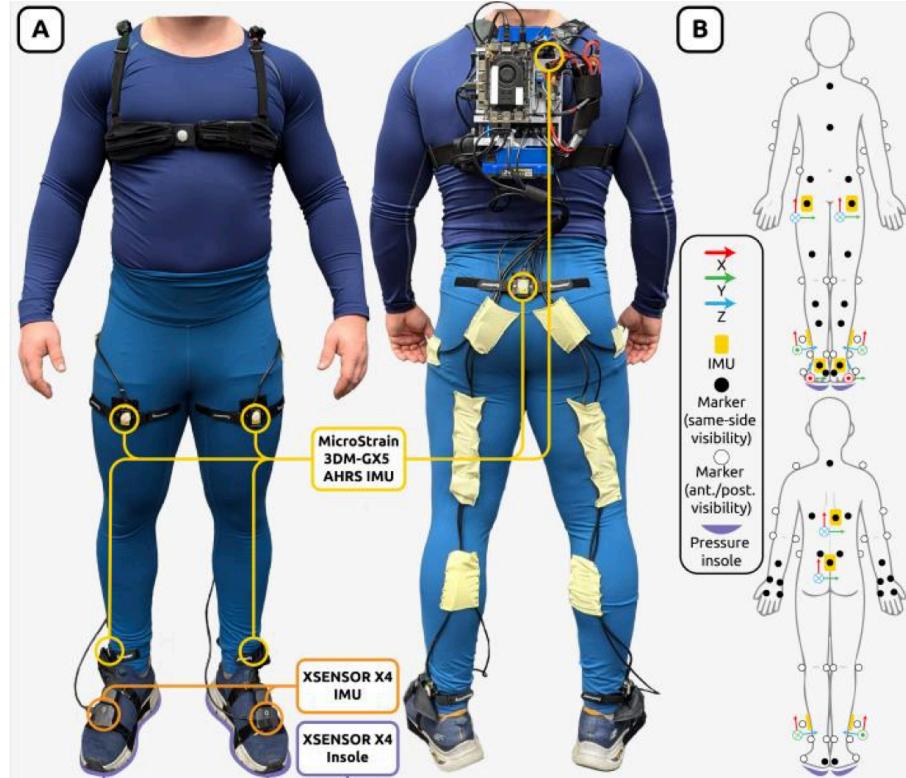
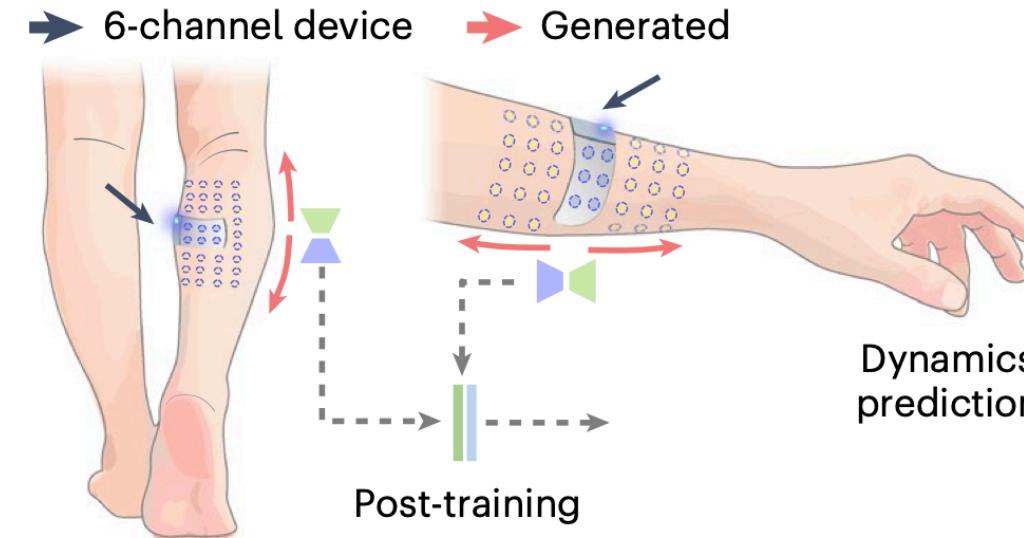


IMU      X-axis  
EMG      Y-axis  
          Z-axis

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into page



# Future Vision: From Rigid Sensors to Soft Biointerfaces

Current (My Work)	Future (Bao Lab Synergy)
 <ul data-bbox="1336 953 2462 1319" style="list-style-type: none"><li>- Rigid IMU sensors</li><li>- Strap-based attachment</li><li>- Fixed sensor locations</li><li>- Rehabilitation-focused</li></ul>	 <ul data-bbox="3691 809 4869 1463" style="list-style-type: none"><li>- Soft, skin-integrated sensors</li><li>- Conformal, comfortable adhesion</li><li>- Flexible placement with generative learning</li><li>- Broader health monitoring applications</li></ul>

Inspired by GenENet (Kim & Bao, *Nature Sensors* 2025)

- Generative representation learning achieves 32-channel EMG performance from 6 channels
- Similar approach could enable rich physiological estimation from minimal soft sensors

**Vision:** Combine my experience in real-time physiological estimation and closed-loop control with Bao Lab's soft sensor expertise to create next-generation wearables that are:

- **Comfortable** enough for all-day wear
- **Intelligent** enough to interpret complex physiology
- **Adaptive** enough to provide personalized interventions

**Applications:** Stroke rehabilitation, Parkinson's monitoring, fall prediction, cardiac health

# Summary & Publications

## Summary

- Developed deep learning methods for real-time joint moment estimation from wearable sensors
- Validated tuning-free hip exoskeleton control in stroke clinical trials
- Designed exoskeleton hardware optimized for clinical populations
- Building open-source multimodal stroke biomechanics dataset

## Selected Publications

### In Preparation

- Park et al. "A versatile, tuning-free hip exoskeleton improves real-world mobility in a clinical trial with stroke survivors" Targeting *Nature Medicine* (Feb 2026)
- Park et al. "A comprehensive dataset of lower-limb biomechanics and wearable sensor measurements in stroke survivors during daily living activities and clinical assessments." Targeting *NEJM AI* (Mar 2026)

### Published

- Park et al. "Human-in-the-loop optimization of hip exoskeleton assistance during stair climbing" *IEEE T-BME* 2025 (Featured Article)
- Kang, Molinaro, Park et al. "Online adaptation framework for exoskeleton assistance in stroke" *IEEE T-RO* 2025
- Choi et al. "Overground gait training with wearable robot in children with cerebral palsy" *JAMA Network Open* 2024

Total: 13 published journal papers & 10 abstracts/conference papers



# Thank you

## Dongho Park

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## Thesis Committee



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(Videos and additional details available on website)

## References

- Dr. Aaron Young (Doctoral Advisor) – [aaron.young@me.gatech.edu](mailto:aaron.young@me.gatech.edu)
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