



This work is also supported by the

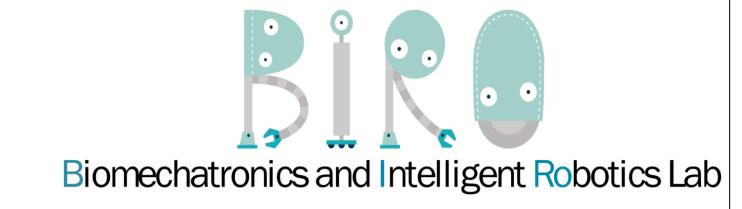
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Learning in Simulation for Exoskeleton-Assisted

Versatile Walking in Community Settings

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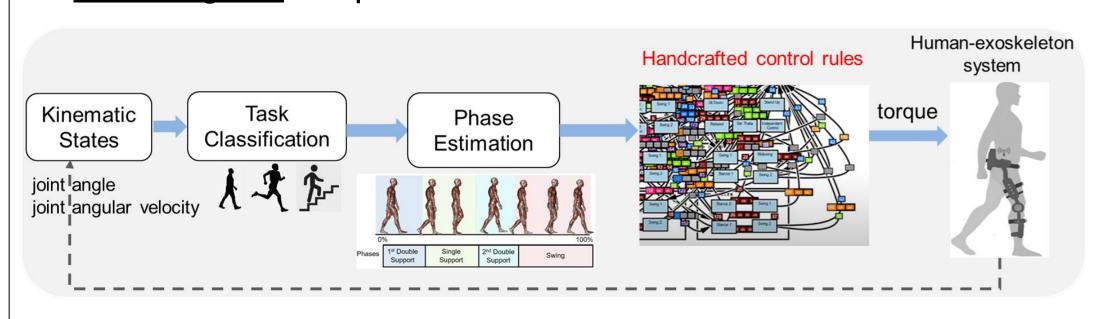




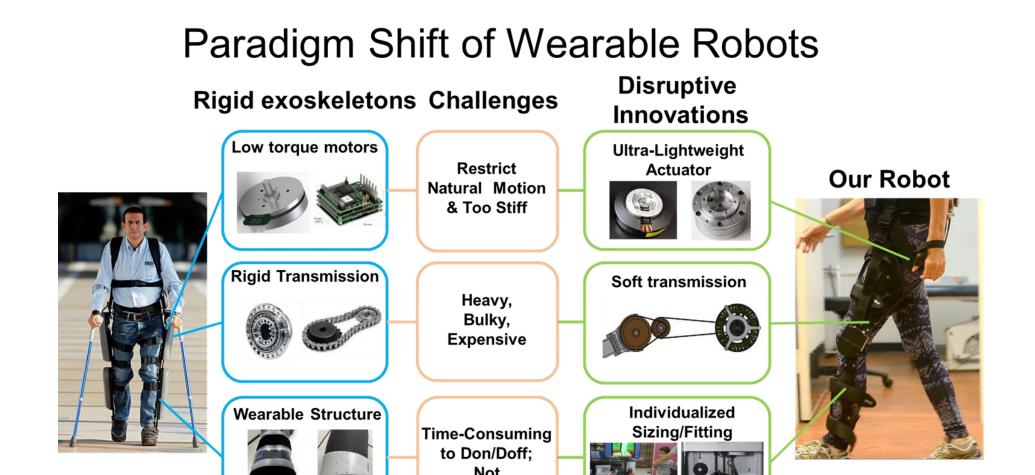


Objectives and Challenges

- Wearable robots like lower-limb exoskeletons have great potential for mobility restoration and human augmentation
- Challenge 1: Required intensive human testing
- Challenge 2: Required handcrafted control laws



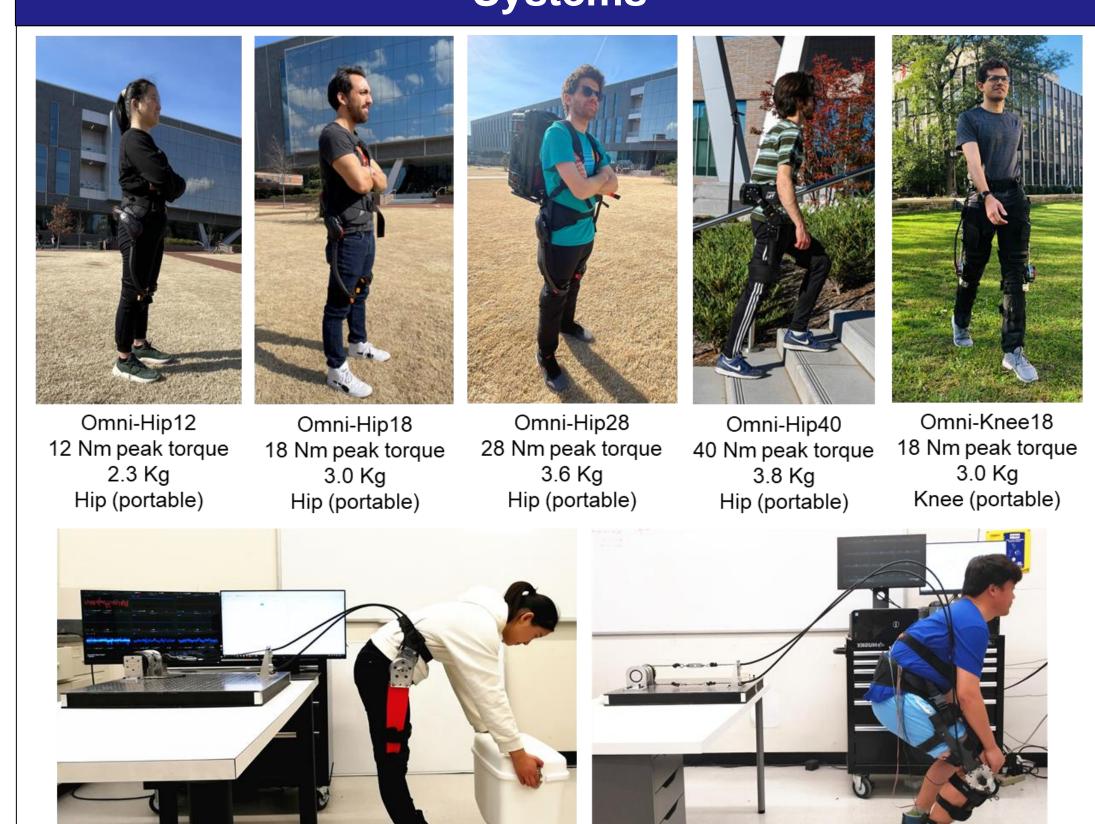
Our Lightweight and High Torque Soft Exoskeleton



Advantages of Our Soft Exoskeleton



Our Portable and Tethered Soft Exoskeleton **Systems**



Omni-Knee72

72 Nm peak torque

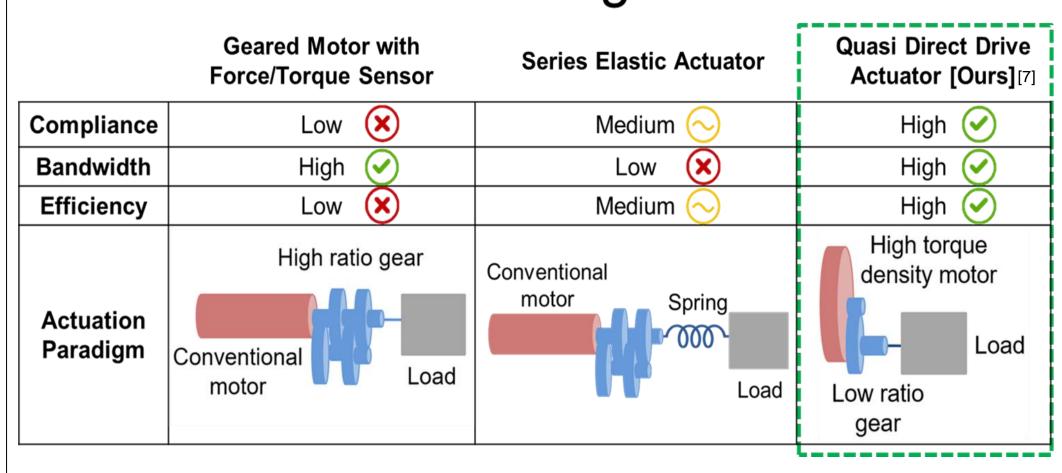
Knee (tethered)

Omni-Hip72

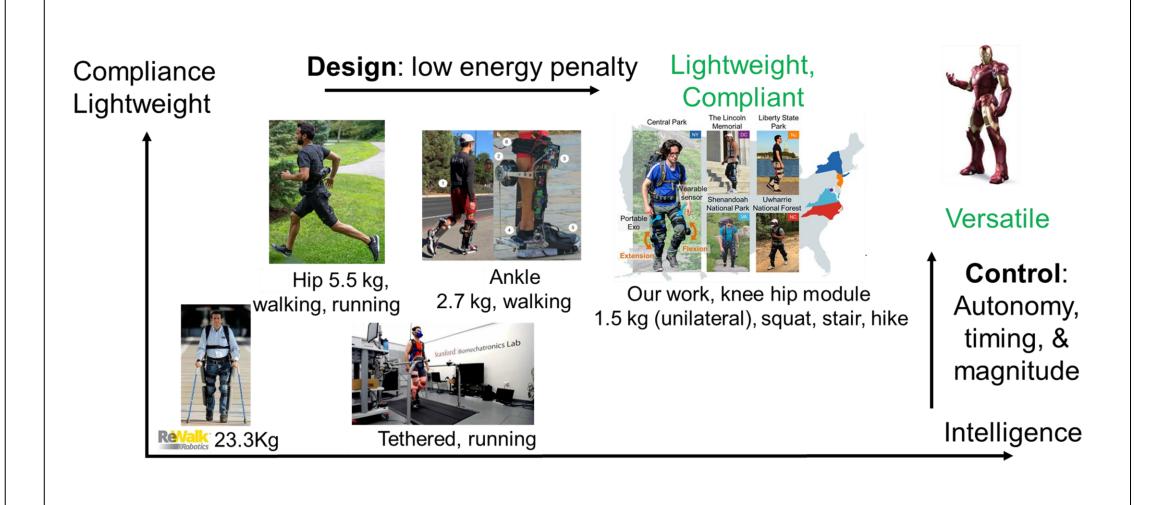
72 Nm peak torque Hip (tethered)

QDD Actuation Paradigm for Exoskeleton

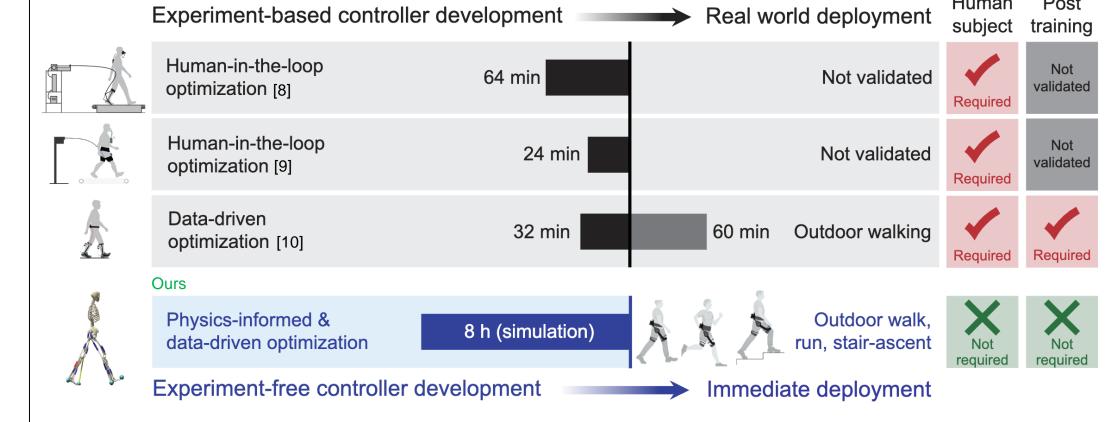
New Actuation Paradigm for Co-Robots



Lightweight, Compliant, and Smart Exoskeletons

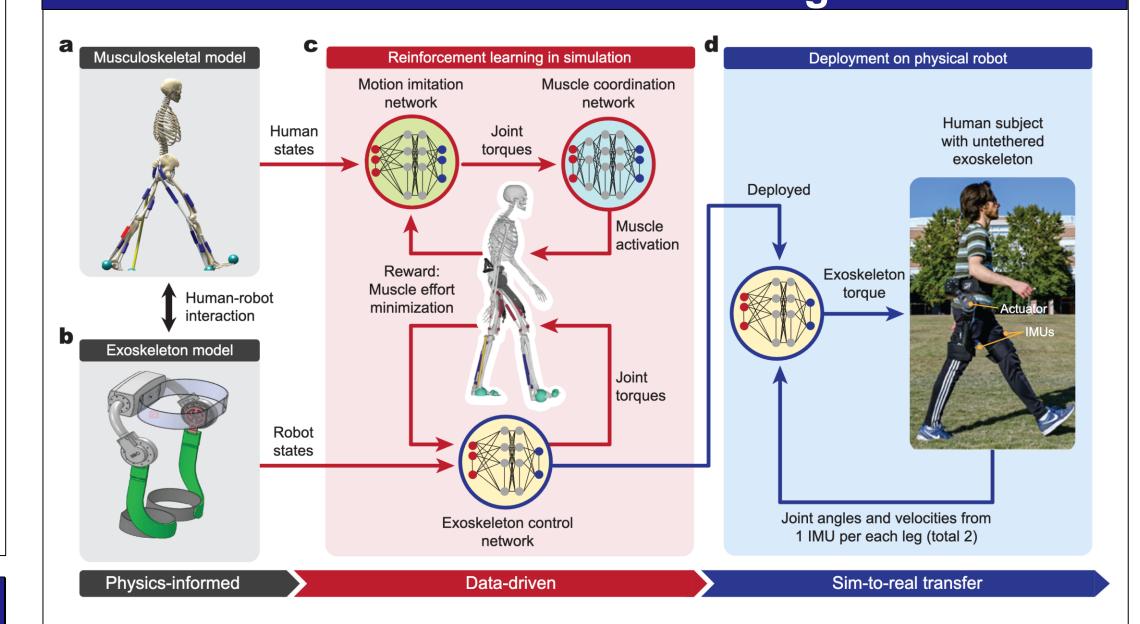


Experiment-free Learning of Exoskeleton Controller In Simulation



- Drawbacks of state-of-the-art methods to get exoskeleton controllers:
 - Requires intensive human experiments for training → This adds formidable cost when applied to another activity or participant
 - Typically for a single activity with steady-state motion → It cannot handle versatile activities or transitions between different activities
- Learning controllers entirely in simulation eliminates the need for human experiments. However, it is still unavailable for wearable robotics community. Key challenges are:
 - Incorporating controller design in the simulation
 - Incorporating human-robot interaction in the simulation
- Our Solution:
 - Eliminates the need for human experiments, learns the exoskeleton controller purely from simulation, and provides immediate energetic benefit to humans
 - Provides synergistic assistance to different subjects for walking, running and stair-climbing

Physics-informed And Data-driven Reinforcement Learning

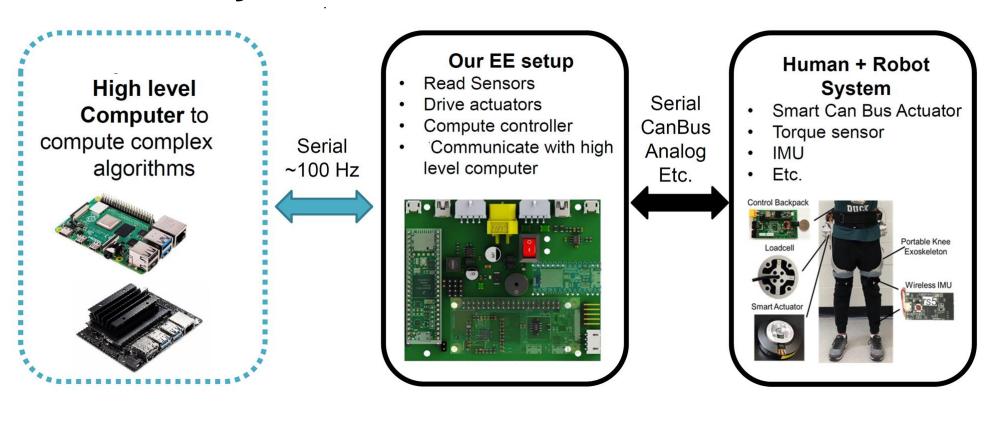


- Our learning method incorporates both physics-informed modeling and data-driven learning:
 - Physics-informed modeling of human musculoskeletal dynamics, exoskeleton, and human-robot interaction
 - Data-driven learning through publicly available human kinematic motion capture dataset
- Our learning method consists of three neural networks that are trained simultaneously for co-evolution:
 - Motion imitation network
 - Muscle coordination network
 - Exoskeleton control network
- Dynamics randomization was used to facilitate Sim-to-real transfer of the trained control policy

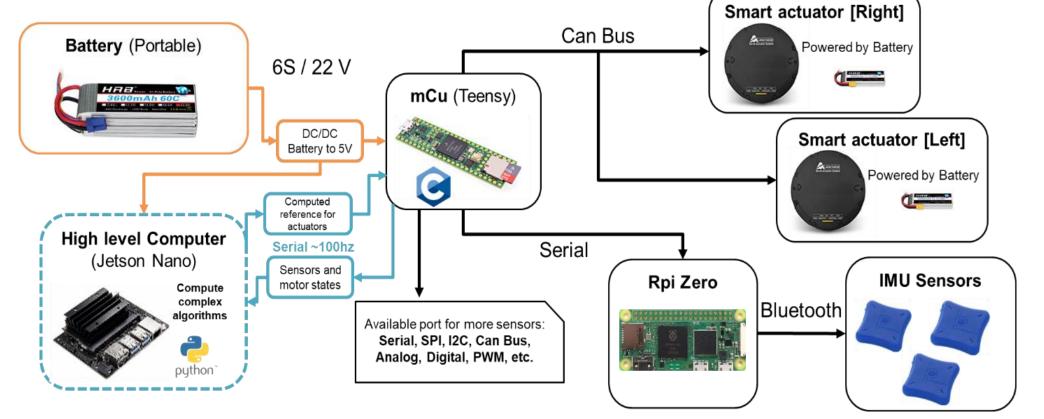
Portable Mechatronics Architecture

 Powerful electronics architecture using a hierarchical structure with a high-level computer and a low-level microcontroller

System Control Architecture

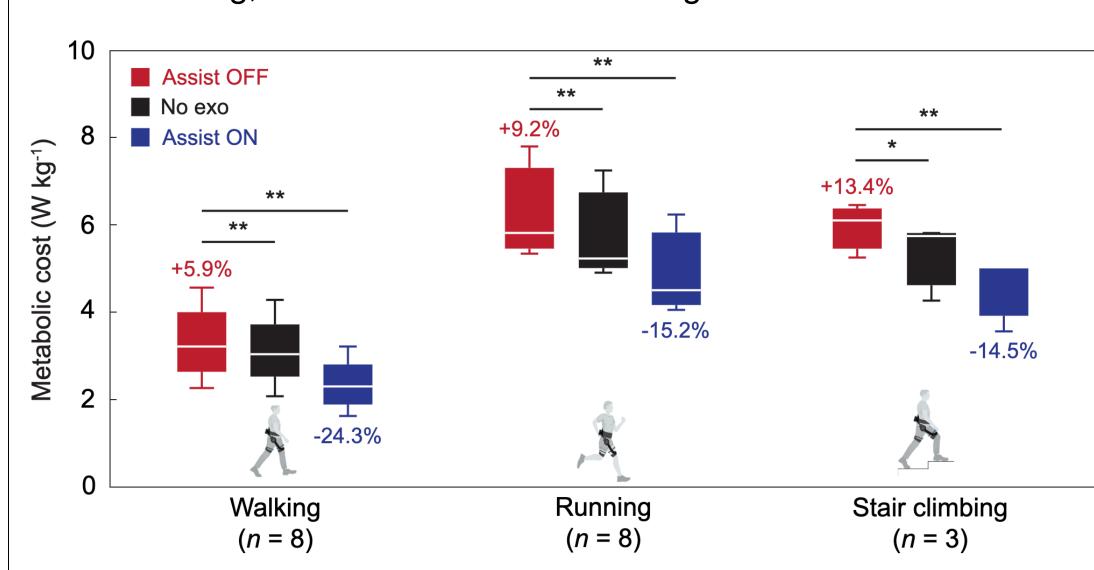


 Able to run complex control algorithms and improve the accuracy, speed, and efficiency of the exoskeleton's control system, leading to better performance, user comfort, and safety

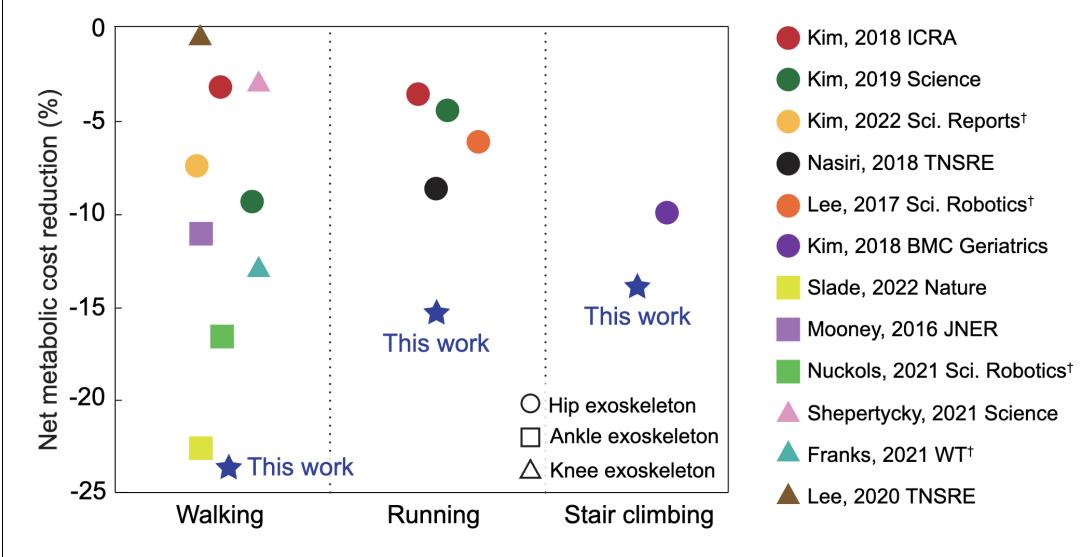


Significant Energetic Cost Reductions on Versatile Activities

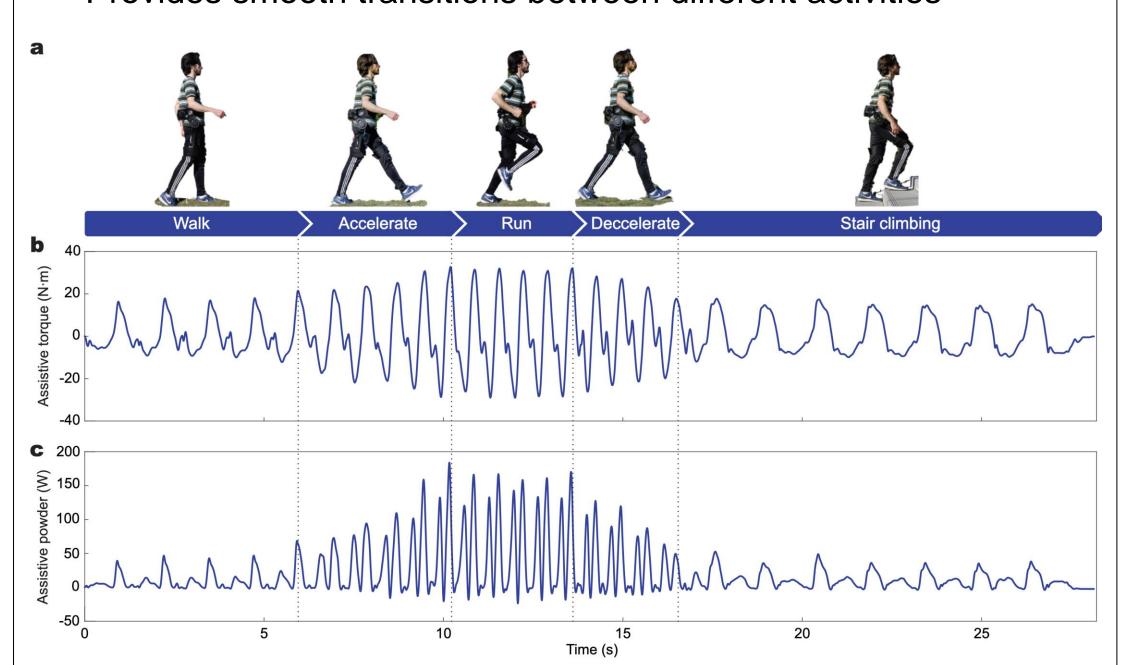
- 8 human subject (5 males, 3 females) experiments utilizing a lightweight, untethered and compliant hip exoskeleton
- Reduced significant metabolic cost by 24.3% for walking, 15.2% for running, and 14.5% for stair climbing



More metabolic cost reduction than state-of-the-art robots



Provides smooth transitions between different activities



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