



Optimal Design of Robust Compliant Actuators for Ubiquitous Co-Robots



NSF Awards
1830360 → 1953908
(Gregg/Rezazadeh)
1830338 (Rouse)

Robert D. Gregg
March 12, 2021

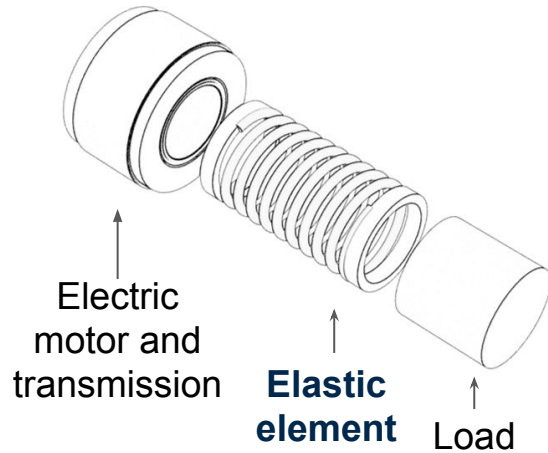
Poster #40

actuation and energy remain major bottlenecks in robotics

G. Yang et al., “*The grand challenges of Science Robotics*,” *Sci. Robot.*, **Jan. 2018**.

A. M. Dollar and H. Herr, “*Lower Extremity Exoskeletons and Active Orthoses: Challenges and State-of-the-Art*,” *IEEE Trans. Robot.*, **Feb. 2008**.

Elasticity!



Series elastic actuator

How to increase
battery life and
quality of force
control?

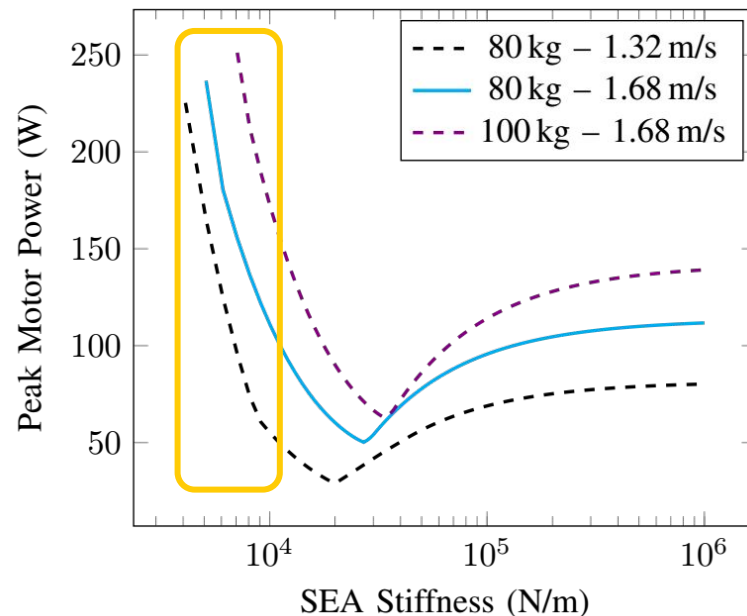
Possible side-effects of SEAs

Poor selection of stiffness for given task may:

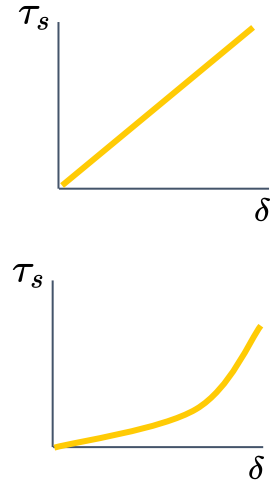
- Increase peak power
- Increase energy consumption

We lack fundamental understanding for robust design!

Example case for a **powered prosthetic ankle**

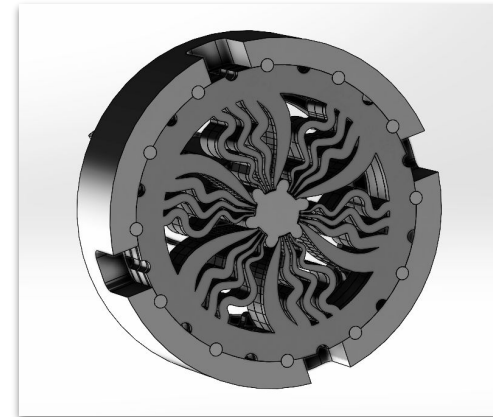


1. Design of **energy efficient series springs** as a **convex** optimization program

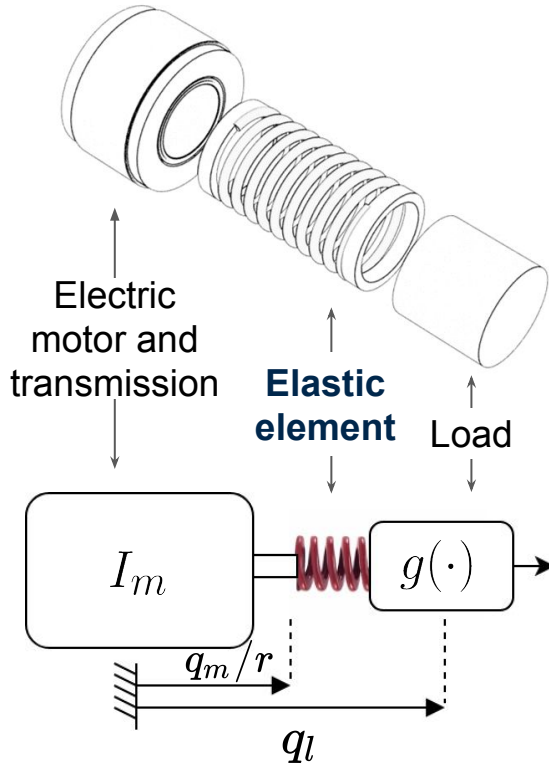


2. **Robust feasible** design of SEAs

3. Mechanical design of **optimal springs**



Modeling of SEAs



Equations of motion for motor shaft:

$$I_m \ddot{q}_m = \tau_m - b_m \dot{q}_m - \frac{\tau_s}{r}$$

Inertia motor & transmission
Motor torque
Viscous friction
Spring torque divided by the ratio of transmission

Spring torque $\tau_s = f(\delta)$

$$\delta = q_l - \frac{q_m}{r}$$

Elongation

Spring torque is equal to load torque:

$$\tau_s = g(q_l, \dot{q}_l, \ddot{q}_l, \tau_e)$$

Known load kinematics and kinetics

Energy consumption

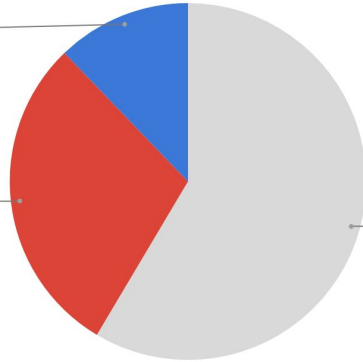


Energy consumption electric motor

Viscous friction
12.2%

Joule heating
29.3%

Load
58.5%



UTD prosthetic leg V1 during walking



Siavash Rezazadeh
U. Denver
(Co-PI)



Edgar Bolívar-Nieto
U. Michigan

$$\begin{aligned}
 E_m &= \int_{t_0}^{t_f} i_m v_m dt = \int_{t_0}^{t_f} \left(\frac{\tau_m^2}{k_m^2} + \tau_m \dot{q}_m \right) dt \\
 &= \int_{t_0}^{t_f} \left(\frac{\tau_m^2}{k_m^2} + \left(I_m \ddot{q}_m + b_m \dot{q}_m + \frac{\tau_s}{r} \right) \dot{q}_m \right) dt \\
 &= \int_{t_0}^{t_f} \left(\underbrace{\frac{\tau_m^2}{k_m^2}}_{\substack{\text{Winding Joule} \\ \text{Heating}}} + \underbrace{b_m \dot{q}_m^2}_{\substack{\text{Viscous} \\ \text{friction}}} - \underbrace{\tau_1 \dot{q}_1}_{\substack{\text{Energy} \\ \text{load}}} \right) dt, \quad \text{where} \\
 & \quad \quad \quad k_m = \frac{k_t}{\sqrt{R}}
 \end{aligned}$$

Energy as a quadratic function of compliance, α

$$E_m = \int_{t_0}^{t_f} \left(\frac{\tau_m^2}{k_m^2} + b_m \dot{q}_m^2 - \tau_l \dot{q}_l \right) dt$$

$$E_m = a\alpha^2 + b\alpha + c$$

Spring Compliance — $\alpha = 1/k$

where

$$a = \int_{t_0}^{t_f} \left(\frac{\gamma_1^2}{k_m^2} + b_m r^2 \dot{\tau}_s^2 \right) dt$$

$$b = \int_{t_0}^{t_f} \left(\frac{2\gamma_1\gamma_2}{k_m^2} - 2b_m r^2 \dot{q}_l \dot{\tau}_s \right) dt$$

$$c = \int_{t_0}^{t_f} \left(\frac{\gamma_2^2}{k_m^2} + b_m \dot{q}_l^2 r^2 - \dot{q}_l \tau_s \right) dt$$

Properties of the quadratic expression

$$E_m = a\alpha^2 + b\alpha + c$$

The quadratic expression is **convex**:

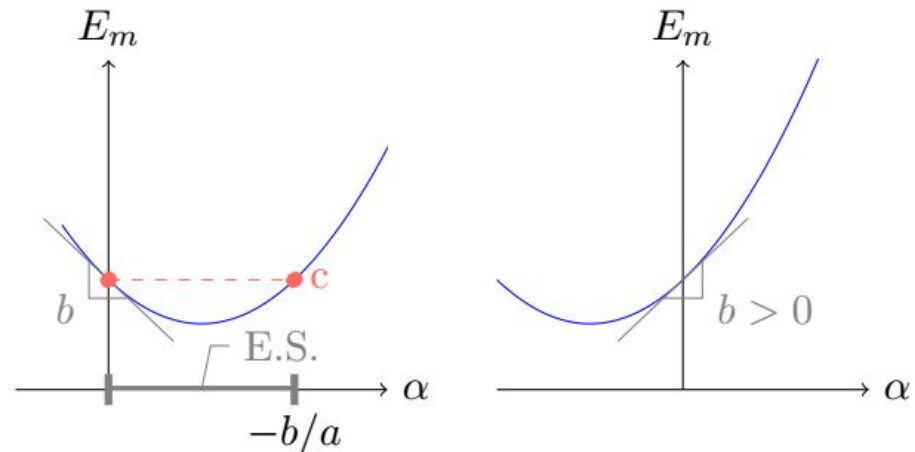
$$\frac{d^2 E_m}{d\alpha^2} = 2a \geq 0$$

C is the energy consumption of a *rigid* actuator:

$$\lim_{k \rightarrow \infty} E_m = c.$$

The **necessary** condition for a linear series elastic element to **save energy**: $b < 0$

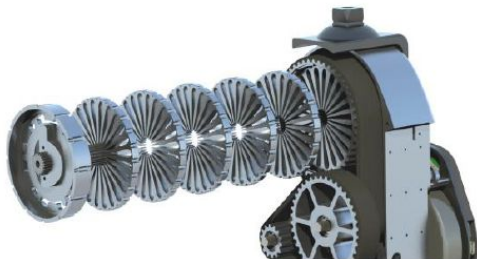
$$\int_{t_0}^{t_f} \left(\frac{2\gamma_1\gamma_2}{k_m^2} - 2b_m r^2 \dot{q}_l \dot{\tau}_s \right) dt < 0$$



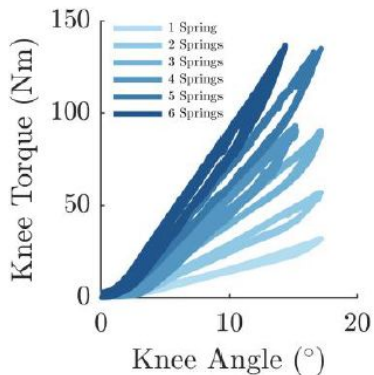
Experimental validation



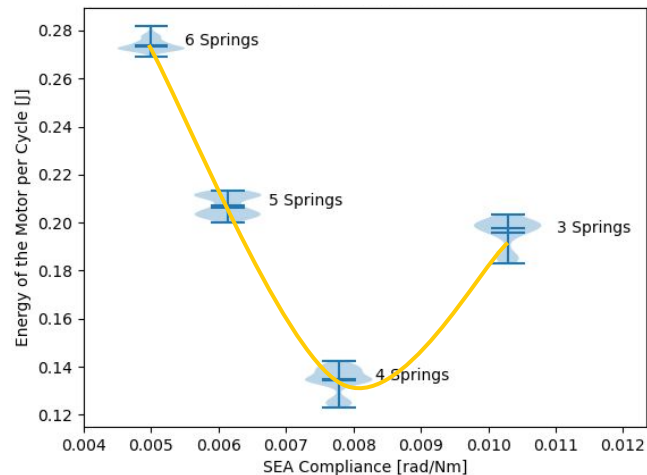
Open Source Leg
U. Michigan



Elliott Rouse
(PI)



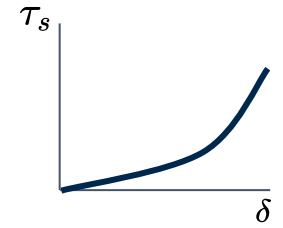
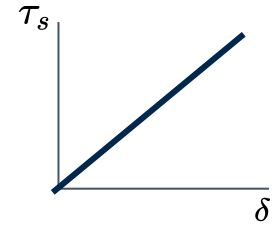
Gray Thomas



$$\begin{aligned} & \underset{\alpha}{\text{minimize}} && \alpha^T G \alpha + h \alpha + w, \\ & \text{subject to} && M \alpha \leq p \end{aligned}$$

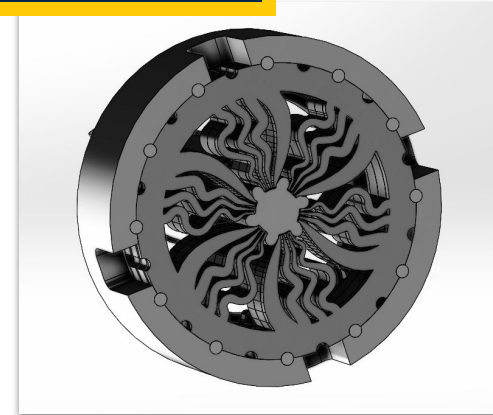
Design of **nonlinear**
and linear springs
that **globally minimize**
energy consumption

1. Design of **energy efficient series springs** as a **convex** optimization program



2. **Robust feasible** design of SEAs

3. Mechanical design of **optimal springs**



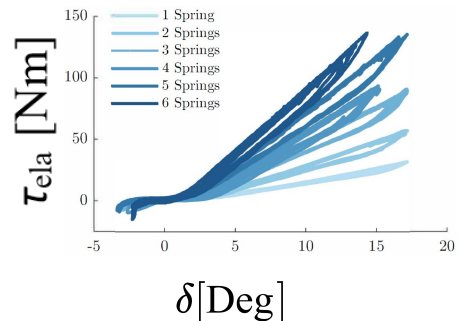
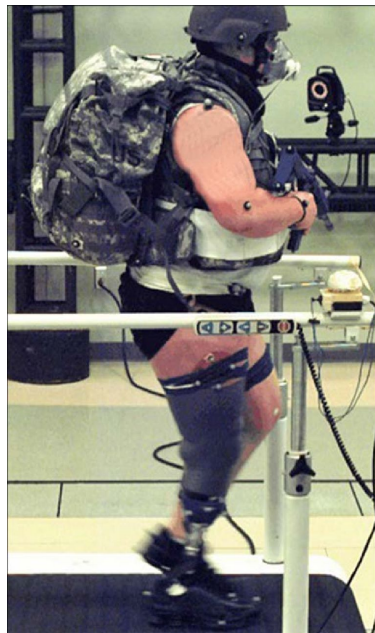
What is robust feasible design?

Uncertainty from:

Manufacturing



Task



Leads to uncertainty in the optimization parameters:

$$\underset{x}{\text{minimize}} \quad f(x, \beta)$$

$$\text{subject to} \quad x \in \mathcal{X}(\beta)$$

$$\beta \in \mathcal{U}_\beta \quad \text{Uncertainty set}$$



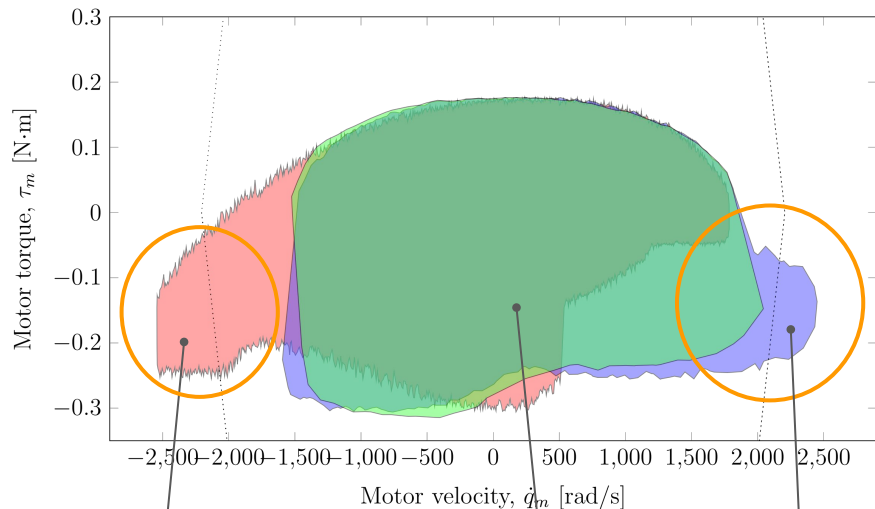
Robust feasible: optimal point is feasible

$$x \in \mathcal{X}(\beta) \quad \forall \beta \in \mathcal{U}_\beta$$

Robust optimal: solution is optimal for the worst case condition

$$\underset{x}{\text{minimize}} \quad \left[\sup_{\beta} f(x, \beta) \right]$$

Motor torque-speed regions for prosthetic ankle



Motor without spring

Robust series spring

Series spring designed with nominal requirements

Uncertain **affine and quadratic** constraints have a computationally tractable counterpart*



Tyler Summers
UT Dallas

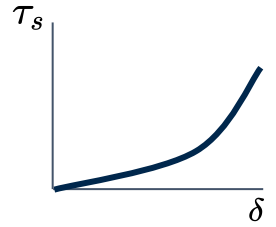
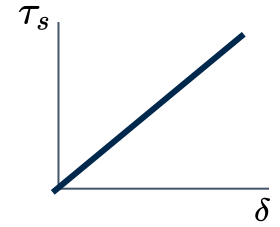


Siavash Rezazadeh
U. Denver



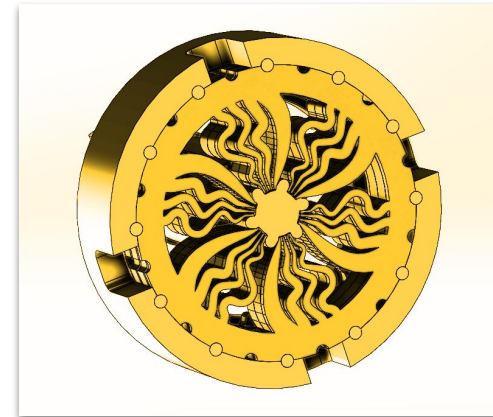
Edgar Bolívar-Nieto
U. Michigan

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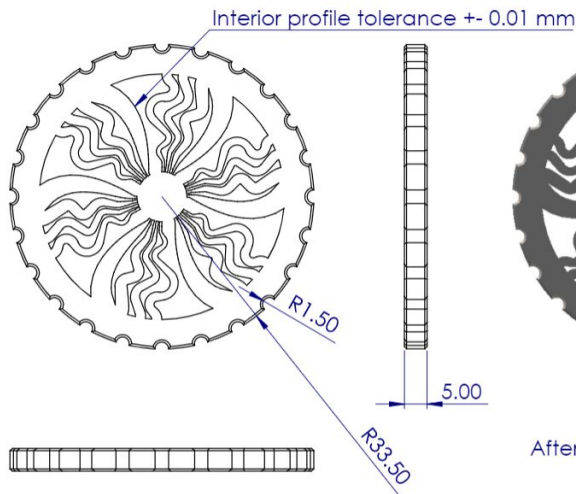
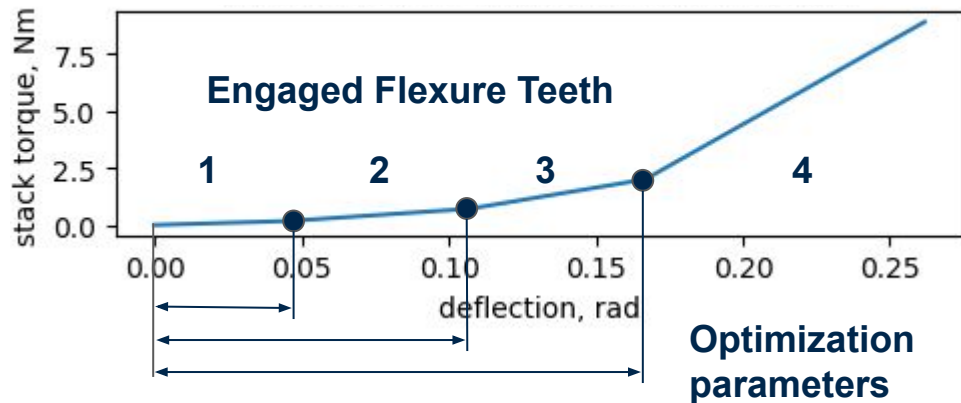


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Non-Linear Disk Spring Design



Gray Thomas
U. Michigan

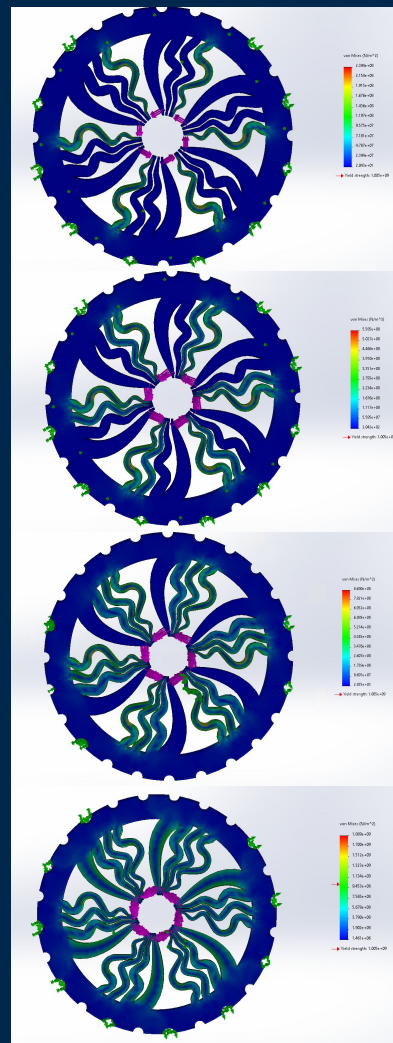
After hardening, temper at 343 C for 2 hours

1st flexure,
until 0.05 rad

Collision with
2nd flexure

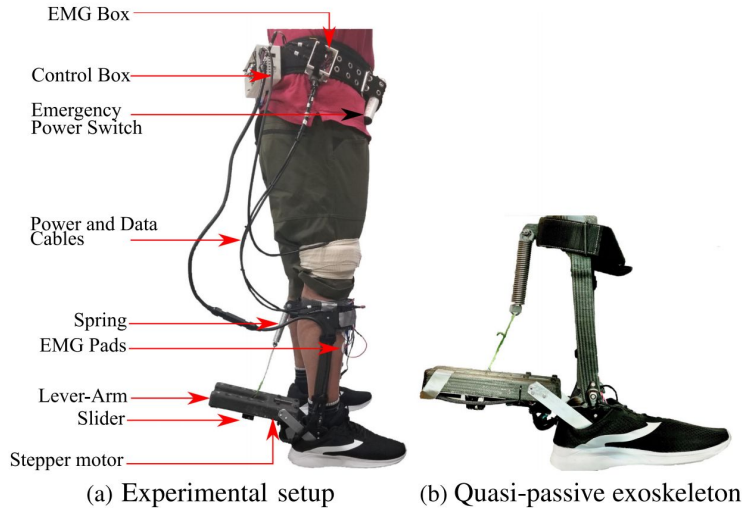
...

At peak loads,
all flexures are
engaged



Broader Impacts

Senior design project led to *RAL* paper!



Control and Optimal Design of Wearable Robots

Summer school 2020

Universidad de los Andes



PhD student graduated



Edgar Bolívar-Nieto, PhD
UT Dallas → U. Michigan

Postdoc supported



Gray Thomas, PhD
U. Michigan

Thank you!



LocoLab and Neurobionics Lab - University of Michigan



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See you at Poster #40!