

HISTORY OF WALKING ROBOTS

Shishir N. Y. Kolathaya

Postdoctoral scholar

Mechanical and Civil Engineering

California Institute of Technology

Seminar

Bombay, Aug 9

Advanced Mechanical
Bipedal Experimental
Robotics LAB

WHY THE DARPA ROBOTICS CHALLENGE TASKS?

The story of the DARPA Robotics Challenge (DRC) begins on March 12, 2011, the day after the Tohoku, Japan earthquake and tsunami struck the Fukushima-Daiichi nuclear power plant. On that day, a team of plant workers set out to enter the darkened reactor buildings and manually vent accumulated hydrogen to the atmosphere. Unfortunately, the vent team soon encountered the maximum level of radiation allowed for humans and had to turn back. In the days that followed, with the vents still closed, hydrogen built up in each of three reactor buildings, fueling explosions that extensively damaged the facility, contaminated the environment and drastically complicated stabilization and remediation of the site.

At Fukushima, having a robot with the ability to open valves to vent the reactor buildings might have made all the difference. But to open a valve, a robot first has to be able to get to it. The DRC tasks test some of the mobility, dexterity, manipulation and perception skills a robot needs to be effective in disaster response.



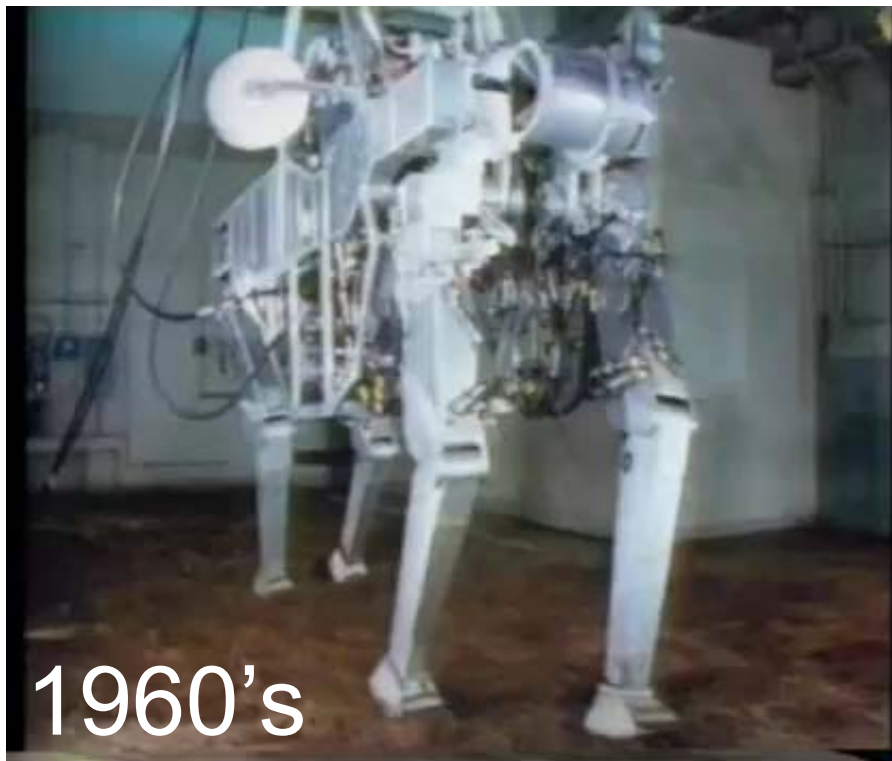
ROBOTICS
CHALLENGE
2013
TRIALS

#DARPADRC

KEY

- Perception
- Mounted Mobility
- Dexterity

- Decision-making
- Dismounted Mobility
- Strength



1960's



1980's



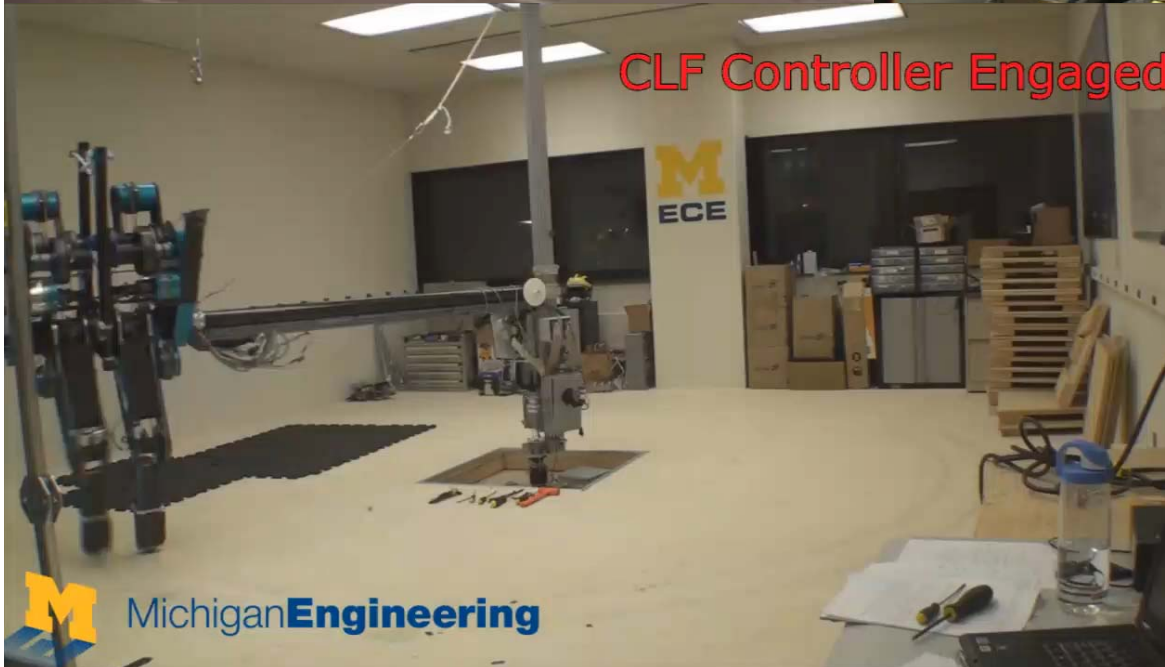
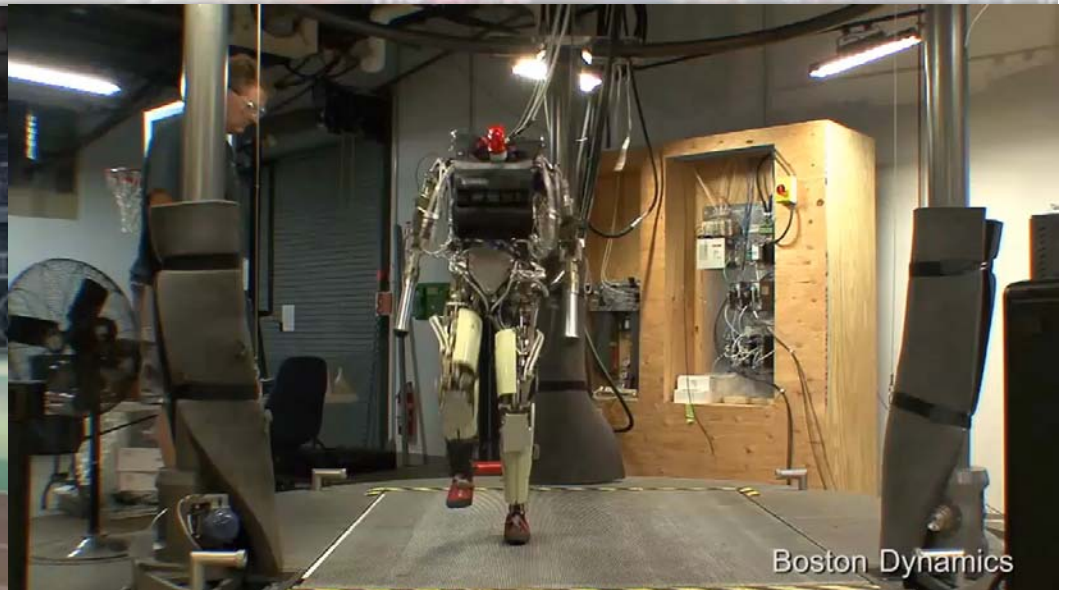
2000's



2000's



There are quite a few working bipeds today!





bipedal robots



All

Videos

Images

News

Shopping

More

Settings

Tools

humanoid robot

exoskeleton

warframe

military

police

scientific

mechanical engineering

japan

simple

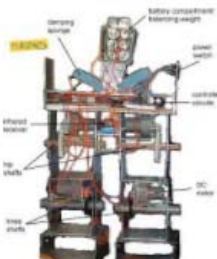
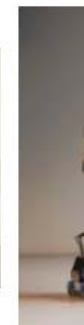
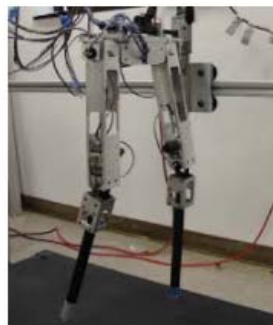
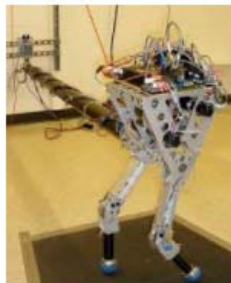
combat

homemade

vex

hector

th



Approaches used for bipedal walking

Traditional Methods:

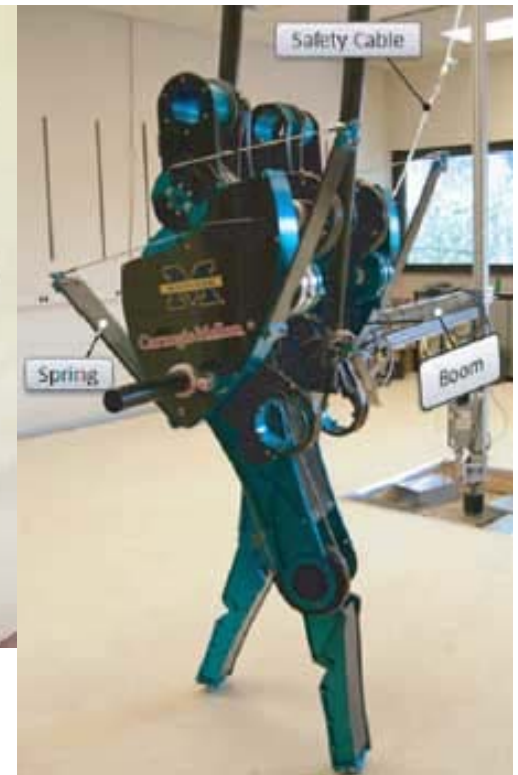
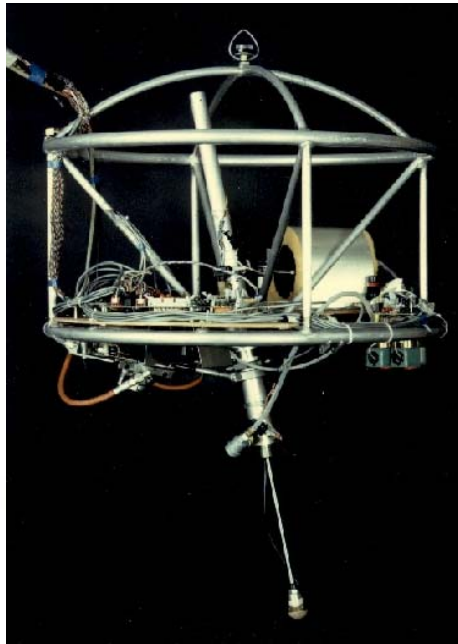
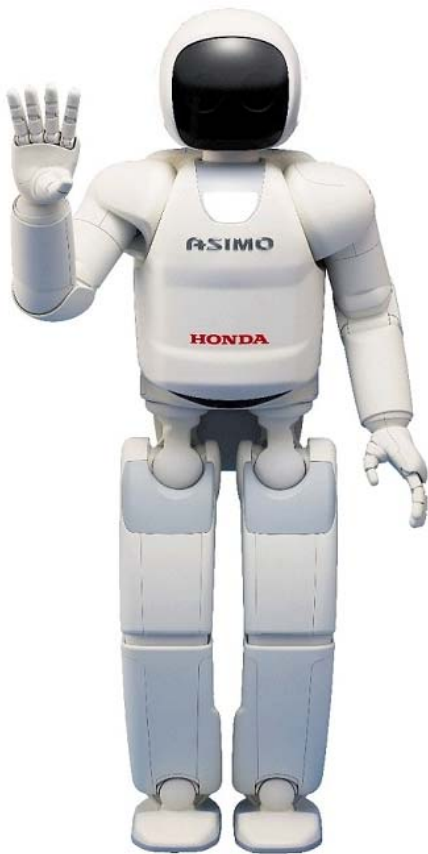
Zero Moment Point (ZMP)

Spring Loaded Inverted Pendulums (SLIP)

Modern Methods:

Passivity based control

Hybrid Zero Dynamics (HZD)



Outline of the talk

History of methods for walking

1. Zero Moment Point (ZMP)
2. Spring Loaded Inverted Pendulums (SLIP)
3. Passivity based control
4. Hybrid zero dynamics (HZD)

Our approaches for walking

1. Gait design:
Direct collocation based trajectory optimization
2. Controller design:
Input-to-state stabilizing controllers

Outline of the talk

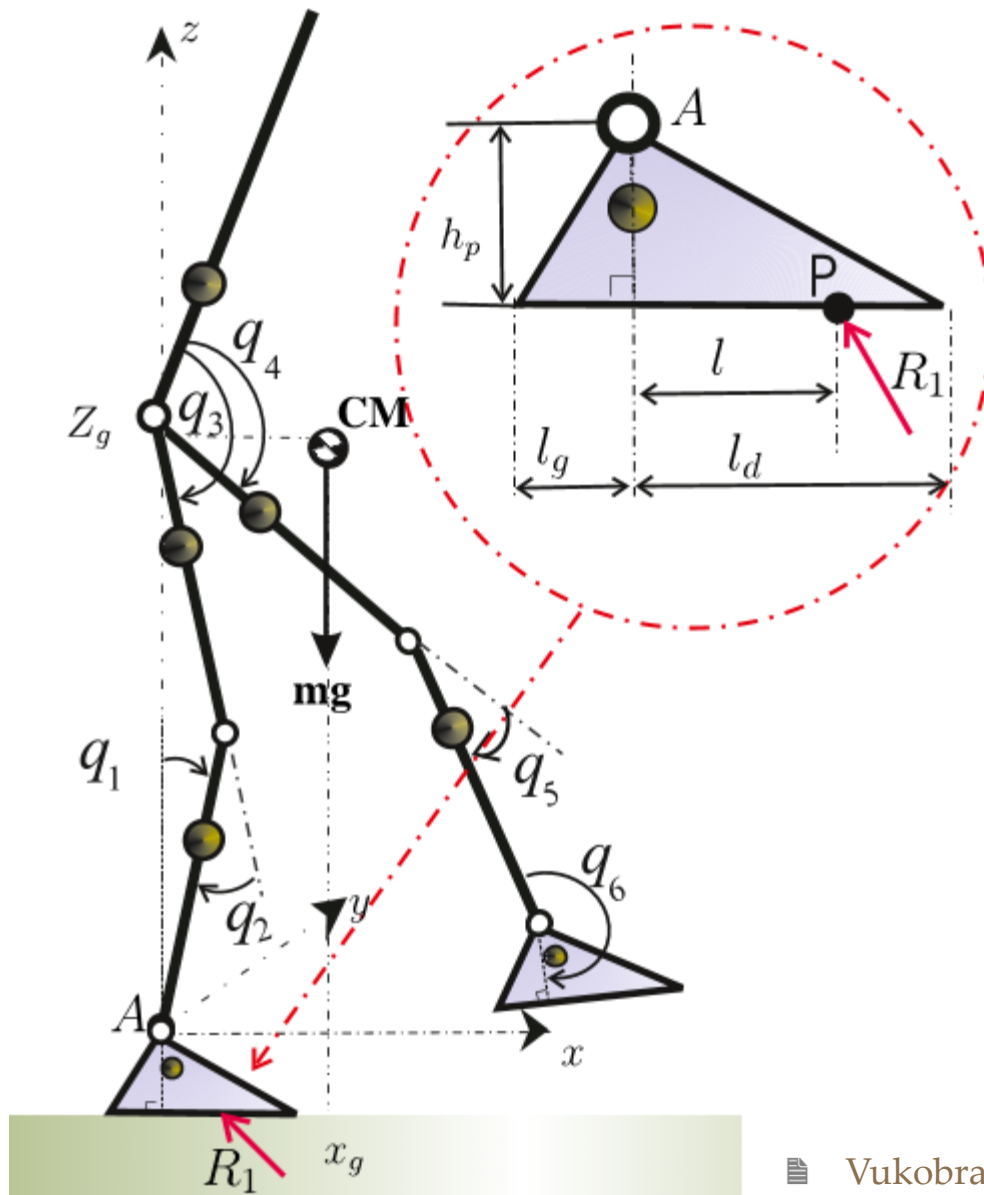
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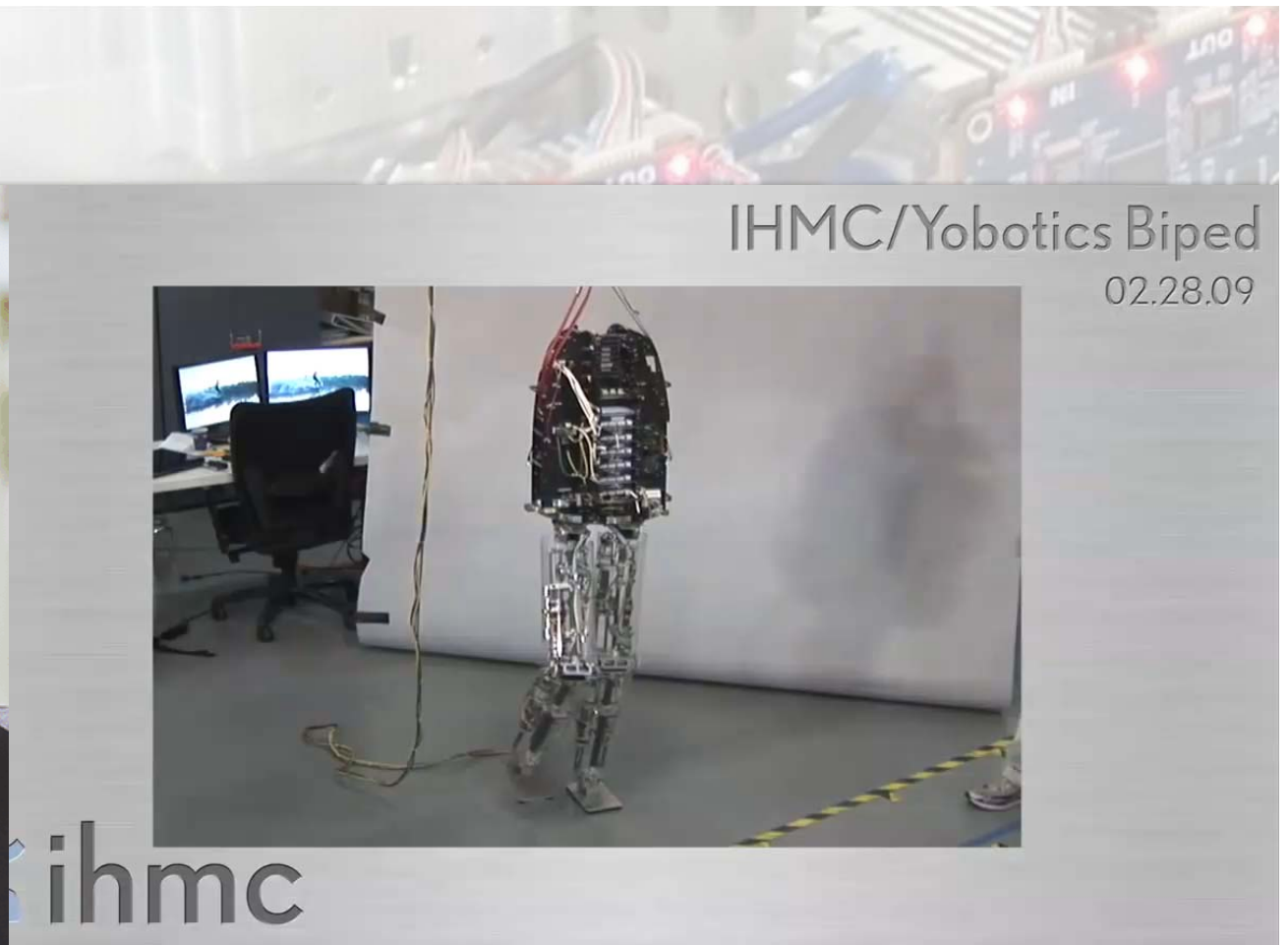
Zero Moment Point (ZMP)



Capture points



Jerry Pratt



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Spring Loaded Inverted Pendulums (SLIP)



Marc Raibert



Outline of the talk

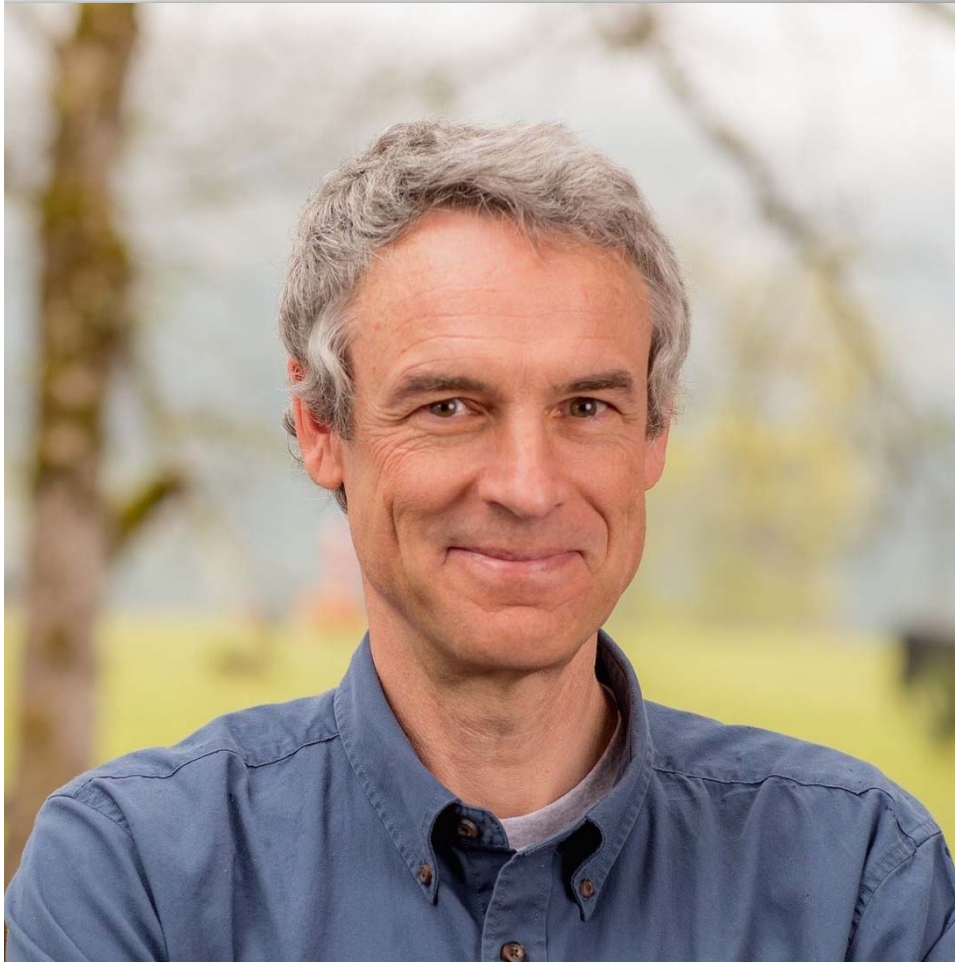
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4. Hybrid zero dynamics (HZD)

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Passivity based control



Tad McGeer

Slope change as a group action

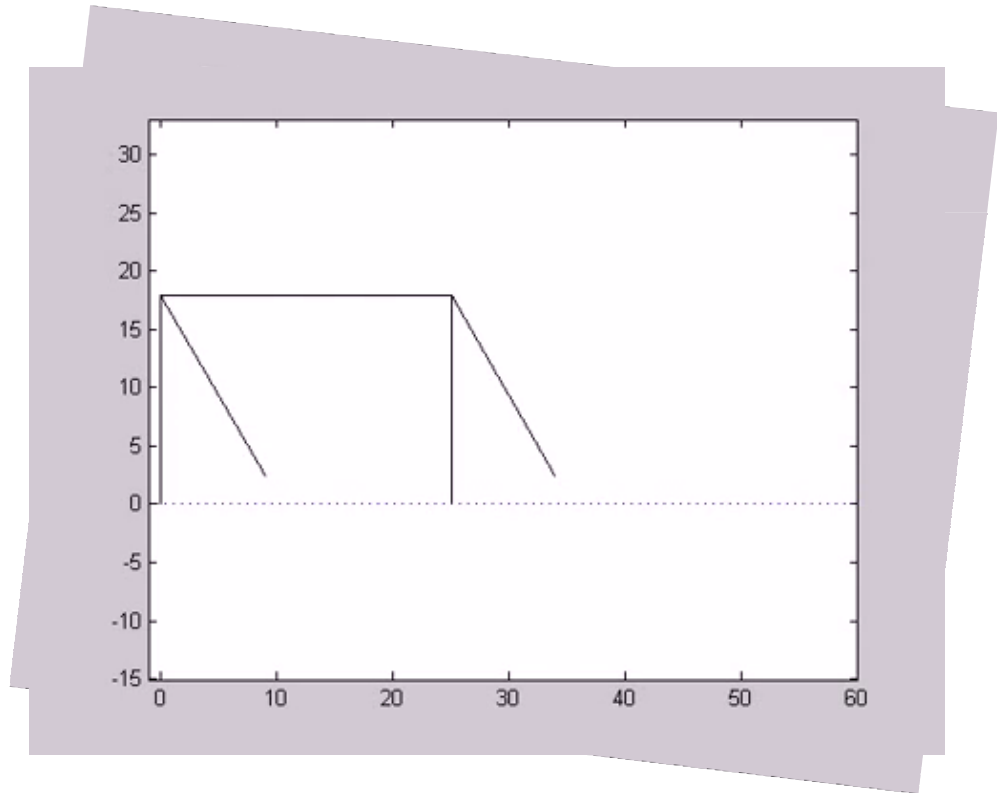
$$Q \subset \mathbb{R}^n, \quad G \subset \mathbb{R}_{\geq 0}$$

θ : Slope angle of the plane

q : Robot configuration

Slope change is a group action

$$\Phi : G \times Q \rightarrow Q$$



The dynamics undergo a controlled symmetrical transformation under the slope change

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Hybrid zero dynamics



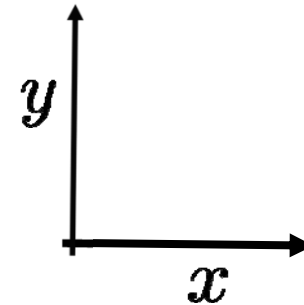
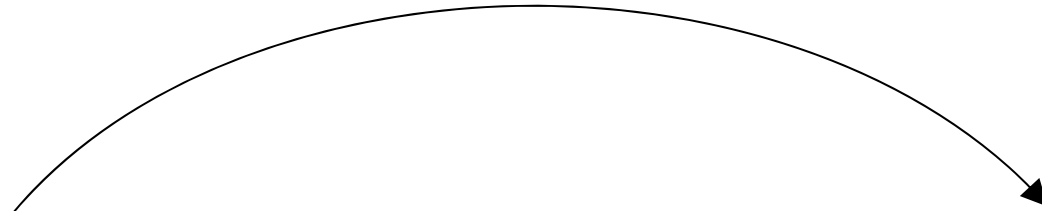
Jessy Grizzle

Zero dynamics \Leftrightarrow Underactuated systems

$$\ddot{x} = \frac{1}{m} F$$

$$\ddot{y} = \frac{1}{m} g$$

$F \rightarrow$

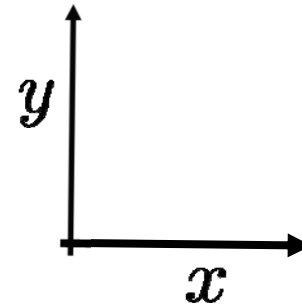
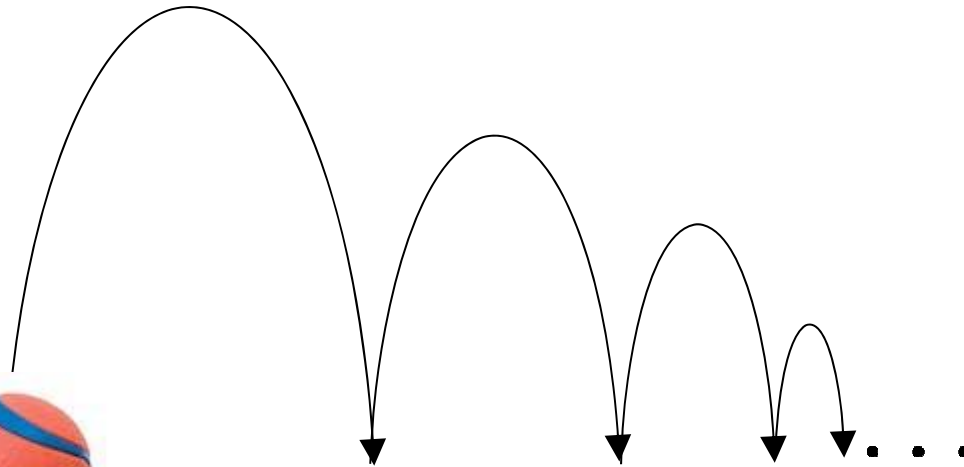


Hybrid zero dynamics \Leftrightarrow Underactuated systems

$$\ddot{x} = \frac{1}{m}F$$

$$\ddot{y} = \frac{1}{m}g$$

$F \rightarrow$



Hybrid zero dynamics \Leftrightarrow Underactuated systems

Tracking control:

Actual outputs and desired outputs:

$y_a(q)$: linear combination of joint angles

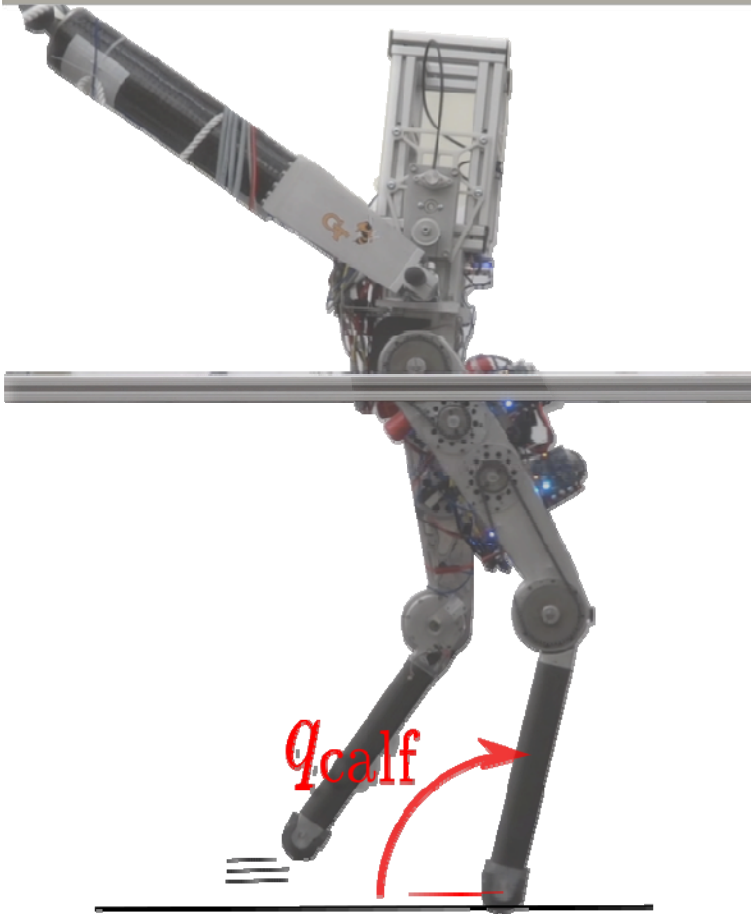
$y_d(q, \alpha)$: desired polynomial trajectory

Example:

$$y_a = \begin{bmatrix} q_{sk} \\ q_{nsk} \\ q_{sh} \\ q_{nsh} \end{bmatrix}$$

$$y_d(q, \alpha) = \begin{bmatrix} \text{polynomial}_1 \\ \text{polynomial}_2 \\ \text{polynomial}_3 \\ \text{polynomial}_4 \end{bmatrix}$$

$$\text{polynomial}_i = \alpha_{i1}t^4 + \alpha_{i2}t^3 + \alpha_{i3}t^2 + \alpha_{i4}t + \alpha_{i5}$$



Hybrid zero dynamics \Leftrightarrow Underactuated systems

Control Objective:

Drive the actual behavior of the robot to the desired behavior:

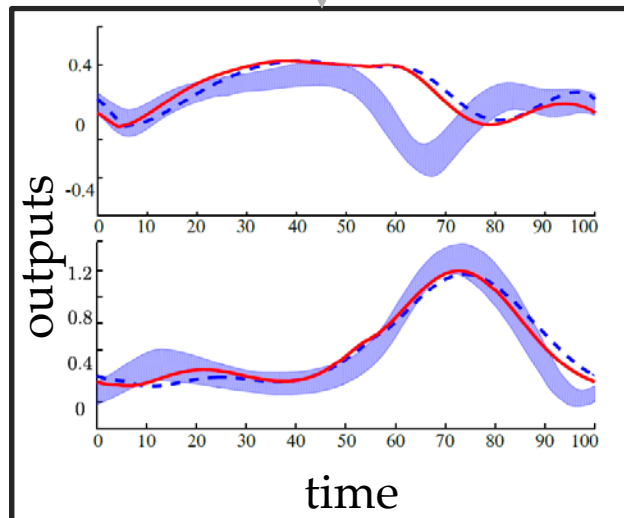
$$y_{\alpha}(q) = y_a(q) - y_d(q, \alpha) \rightarrow 0$$



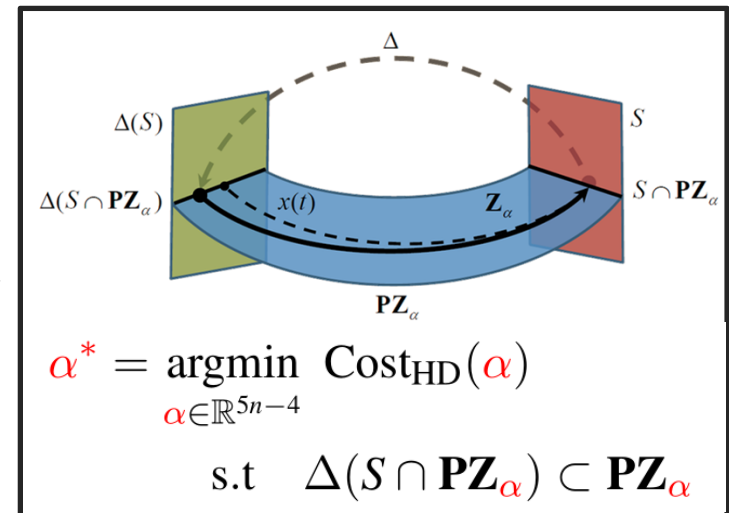
$$y_d(q, \alpha)$$



$$y_a(q)$$



$$y_a(q) \rightarrow y_d(q, \alpha)$$



Hybrid Zero Dynamics

Hybrid zero dynamics \Leftrightarrow Underactuated systems

Tracking control:

Actual outputs and desired outputs:

$y_a(q)$: linear combination of joint angles

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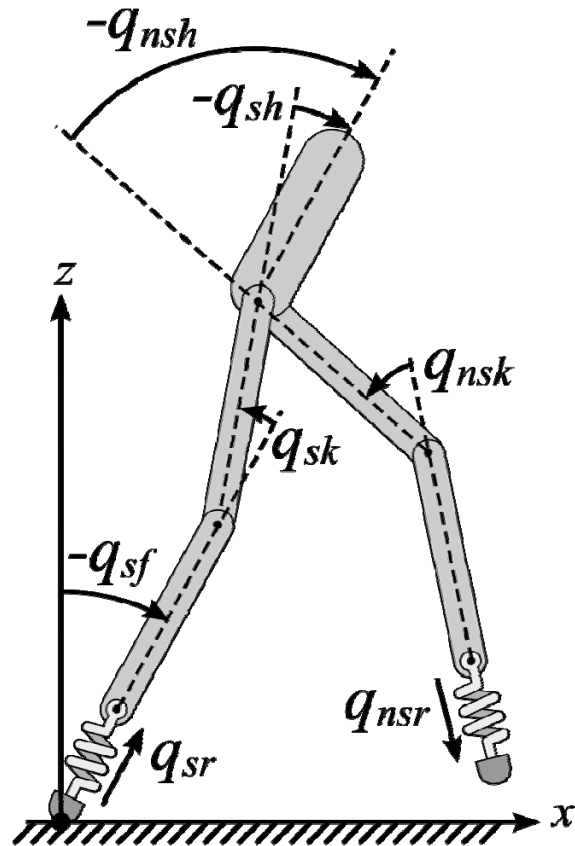
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$$\text{polynomial}_i = \alpha_{i1} t^4 (q_{calf})^4 + \alpha_{i2} t^3 (q_{calf})^3 + \alpha_{i3} t^2 (q_{calf})^2 + \alpha_{i4} t (q_{calf}) + \alpha_{i5}$$

Hybrid zero dynamics \Leftrightarrow Underactuated systems



Tracking control:

Actual outputs and desired outputs:

$y_\alpha(q)$: linear combination of joint angles

$y_d(q, \alpha)$: desired polynomial trajectory

Example:

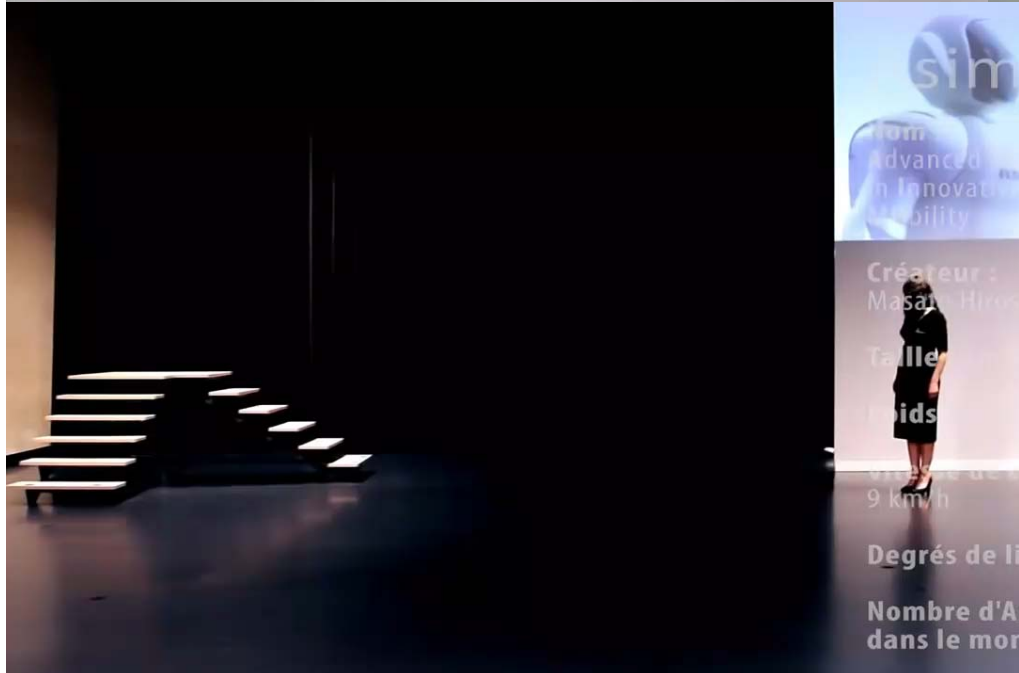
$$\begin{bmatrix} \dot{y}_\alpha \\ \ddot{y}_\alpha \end{bmatrix} = f(y_\alpha, \dot{y}_\alpha) + g(y_\alpha, \dot{y}_\alpha)u \quad \left. \vphantom{\begin{bmatrix} \dot{y}_\alpha \\ \ddot{y}_\alpha \end{bmatrix}} \right\} \text{Output dynamics}$$

$$\dot{z} = \Psi(y_\alpha, \dot{y}_\alpha, z) \quad \left. \vphantom{\dot{z}} \right\} \text{Passive dynamics}$$

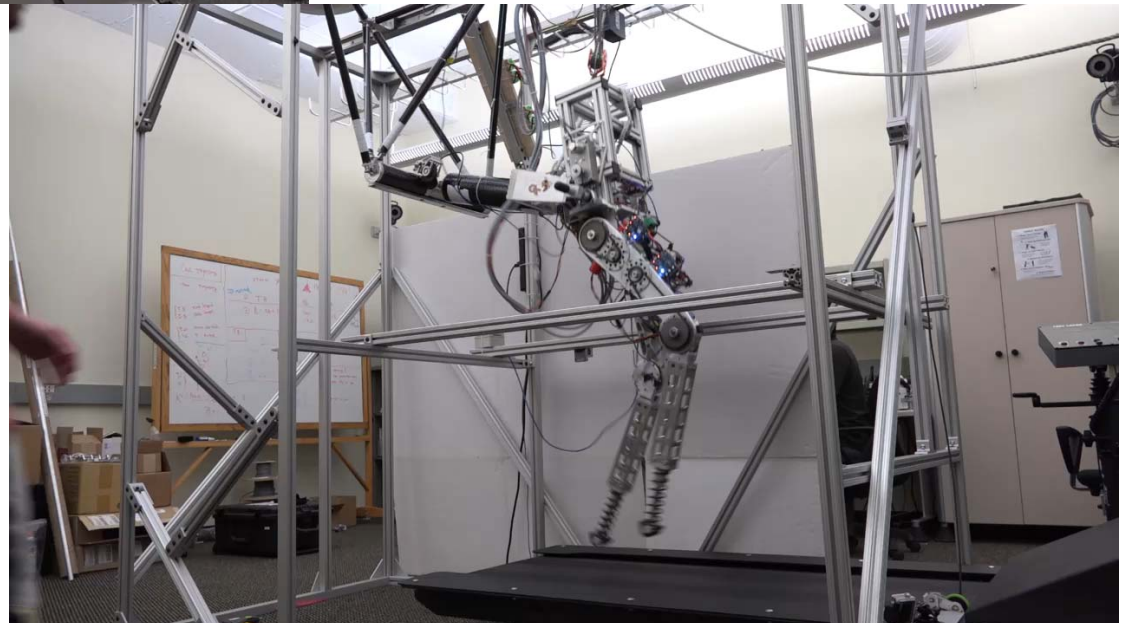
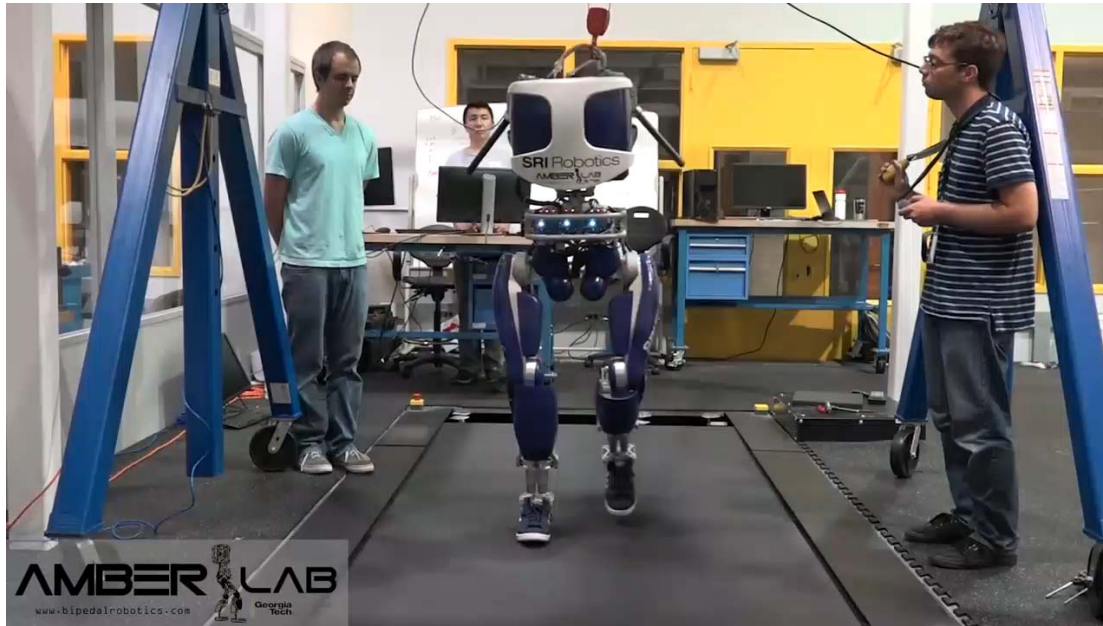
Gait design

Control design

We have progressed from this



To this...



Outline of the talk

History of methods for walking

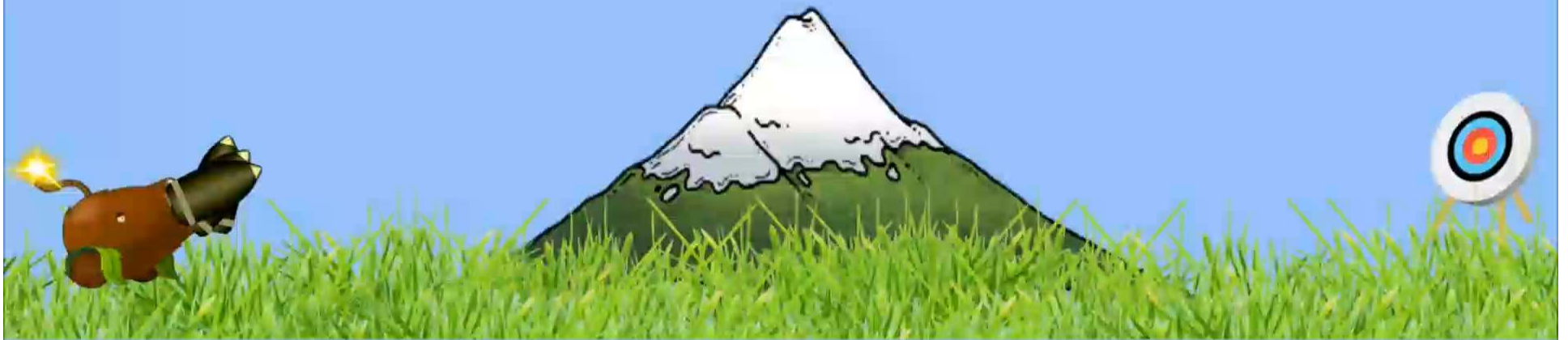
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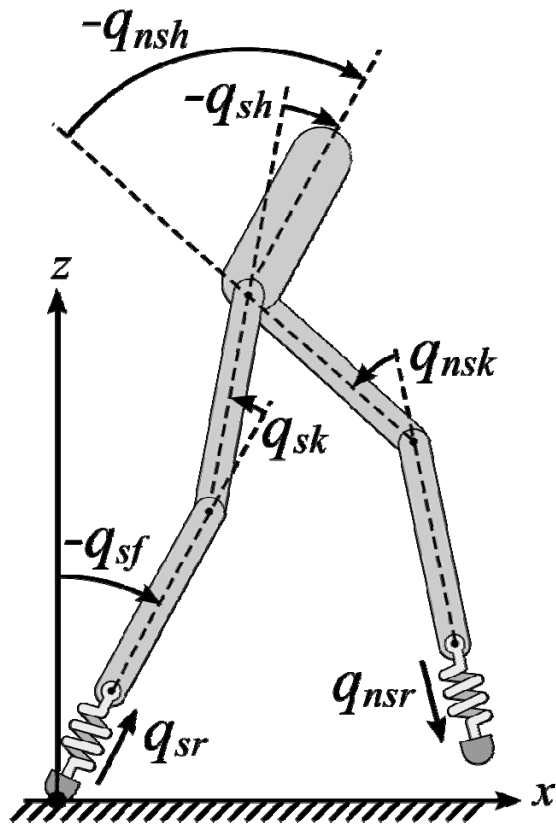
Gait design \Leftrightarrow Trajectory Optimization

Trajectory Generation / **Optimization**



We need fast trajectory optimization: Limit cycle in simulation does not imply limit cycle in experiment.

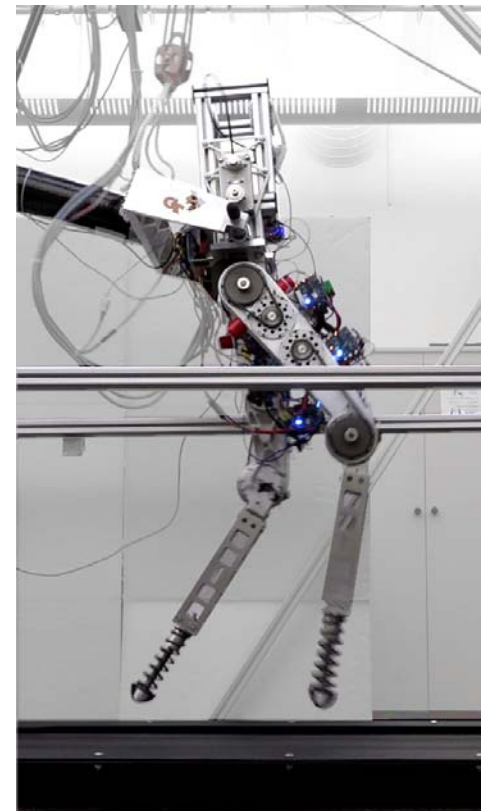
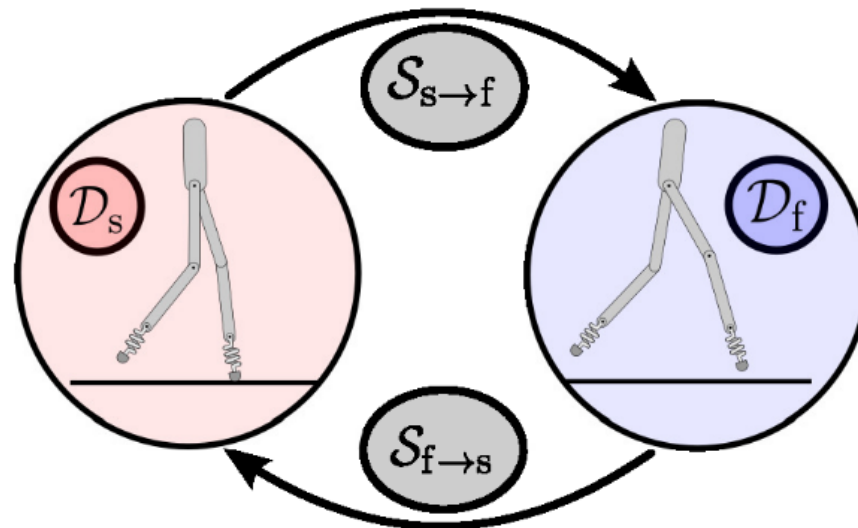
Trajectory Optimization



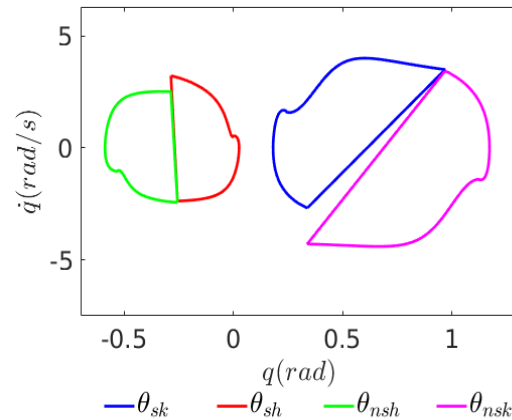
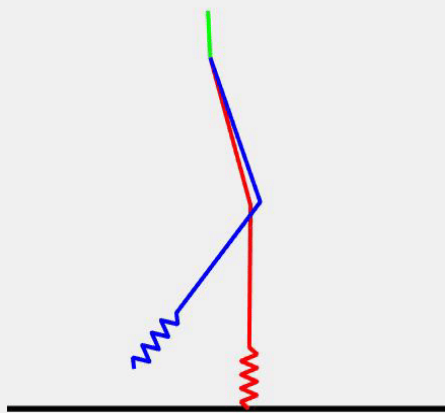
$$q \in \mathbb{R}^n, u \in \mathbb{R}^m, F \in \mathbb{R}^l$$

$$v \in \{s, f\}, \quad e \in \{s \rightarrow f, f \rightarrow s\}$$

$$\{\mathcal{D}_s, \mathcal{D}_f\}, \quad \{\mathcal{S}_{s \rightarrow f}, \mathcal{S}_{f \rightarrow s}\}$$

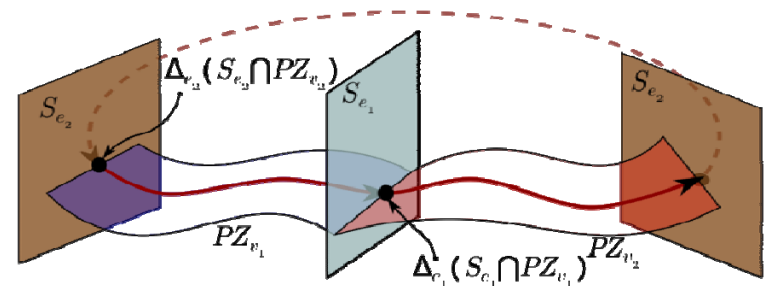


Optimization performance

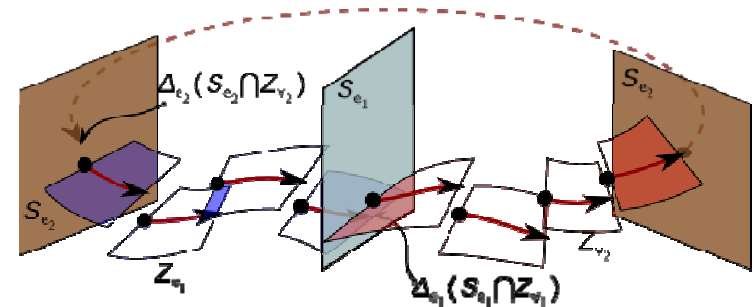


Periodic orbit \mathcal{O}

Single shooting



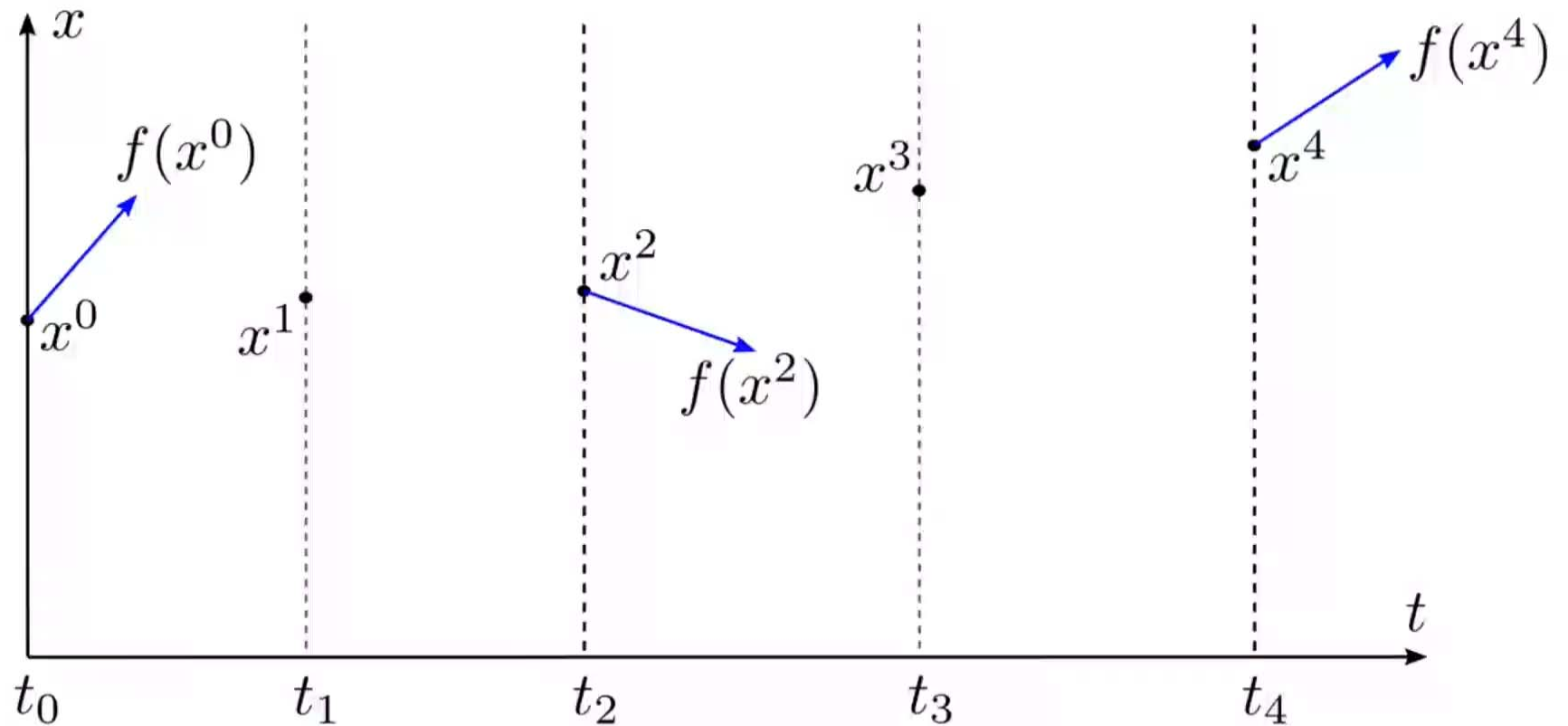
Multiple shooting



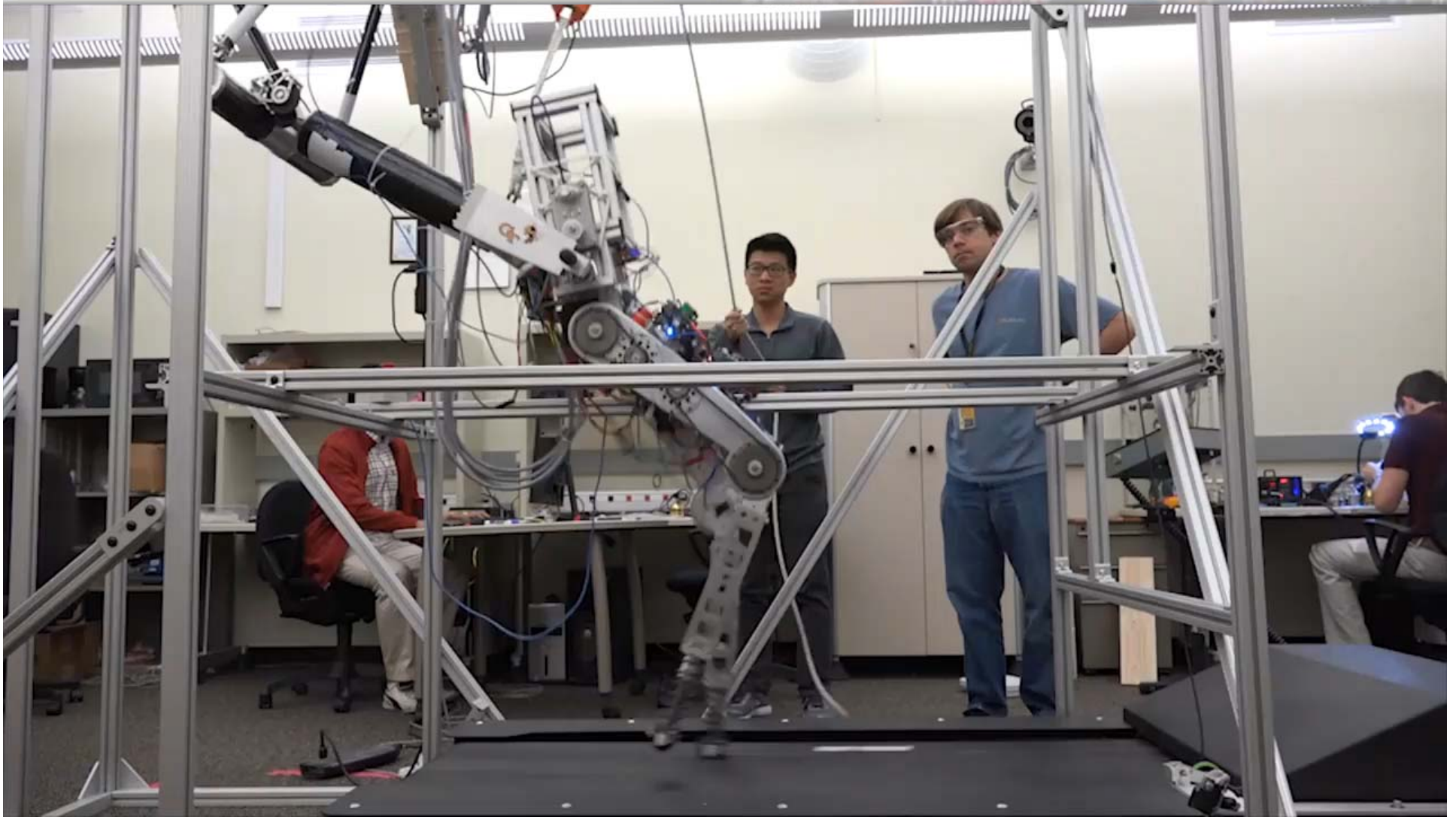
1. Single shooting method: >2 hours, solution not found
2. Multiple shooting method: 13 hours, solution found
3. **Direct collocation: <1 min, solution found**

Trajectory Optimization

Direct Collocation



Trajectory Optimization



We need fast trajectory optimization for testing different gaits.

Outline of the talk

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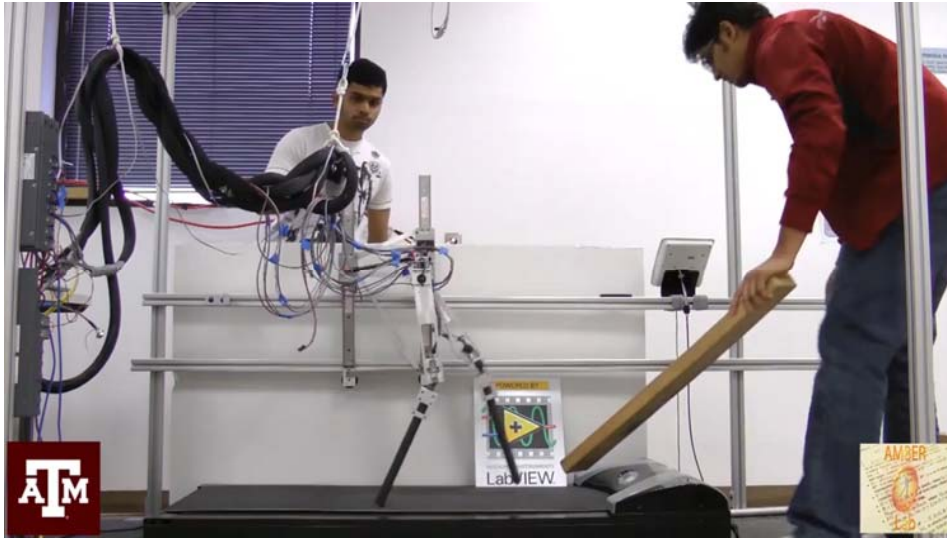
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Controller design plays an important part!

Dec 2011



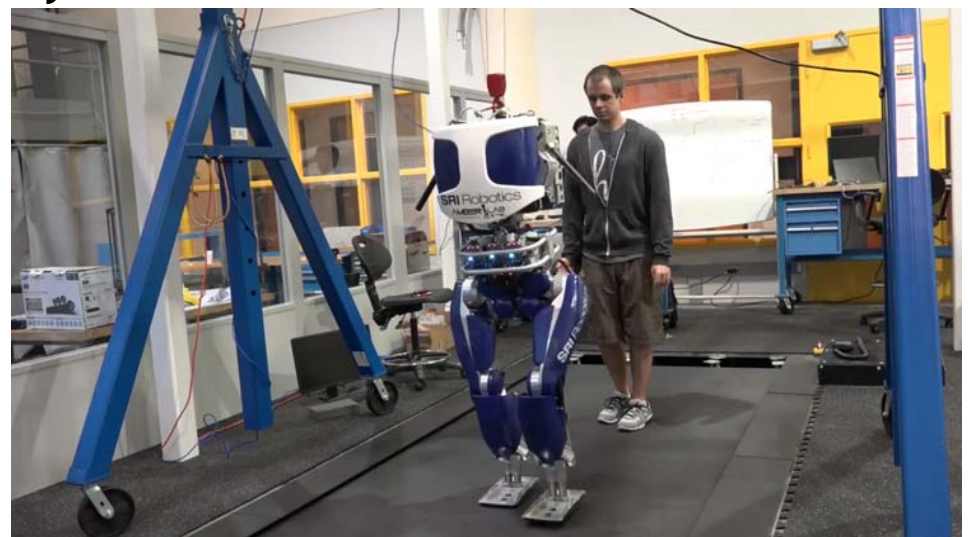
Nov 2013




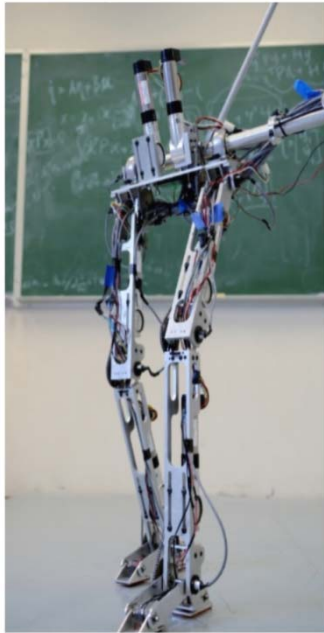
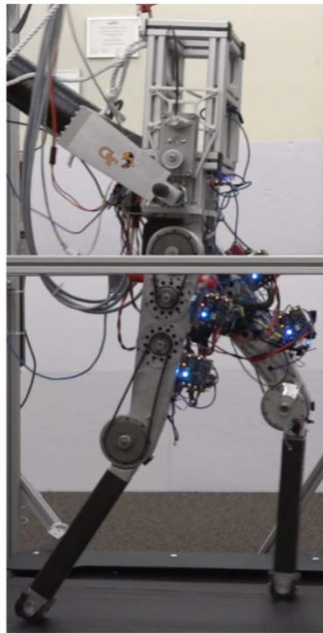
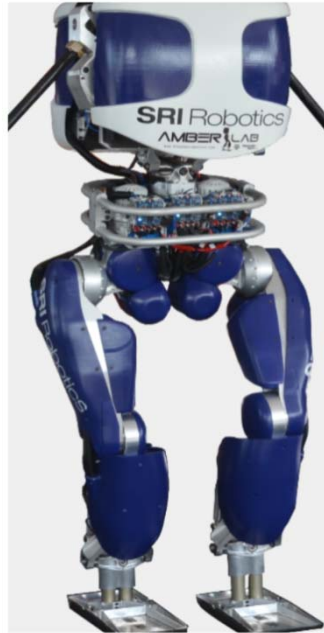
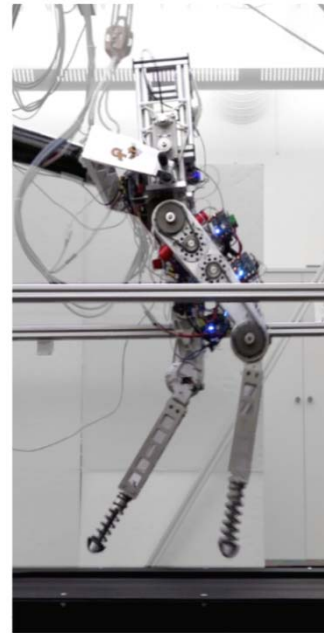
Jun 2015



Jun 2016

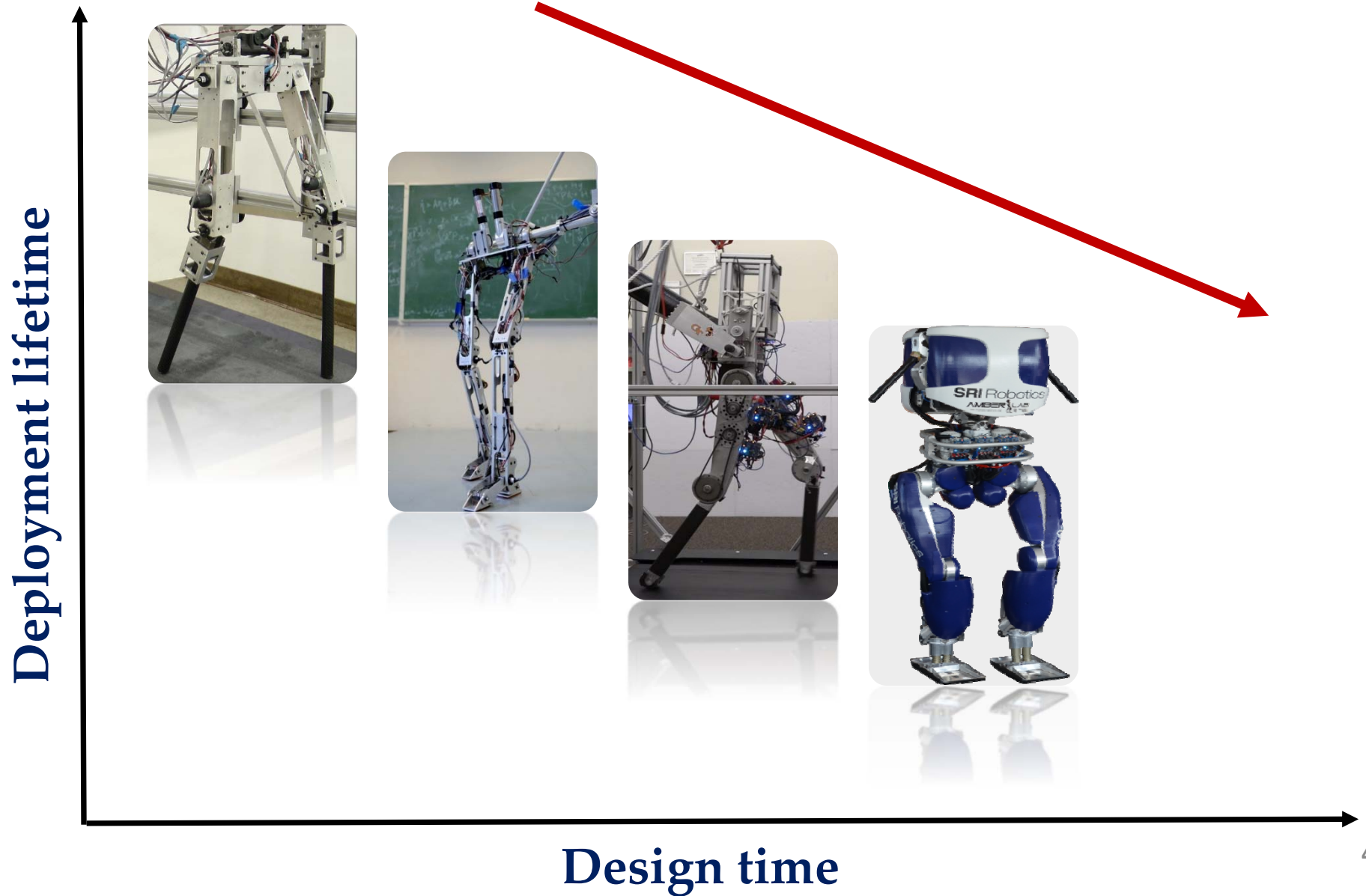


Controller design plays an important part!

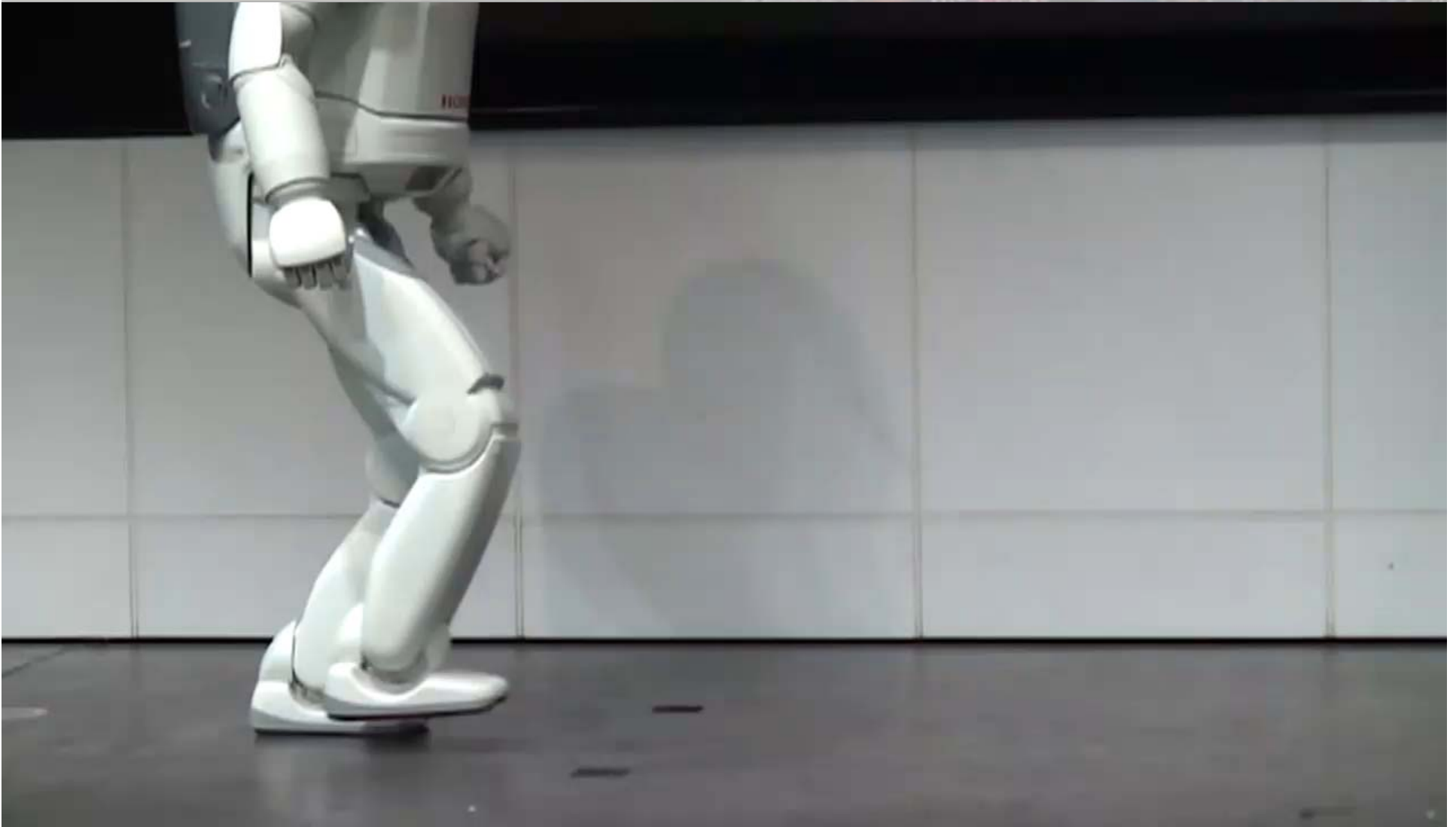
				
AMBER1	AMBER2	PROXI	DURUS	DURUS-2D
P Voltage Control	PD current control	PD current control	PD current control+Regulators, Time based parameterization	PD current control, Time+state based parameterization
2011	2013	2014	2015	2016

➔ **Increasing complexity!**

Controller design plays an important part!



Returning to basics!



Watch how the hip is always maintained at a constant height.

Returning to basics!

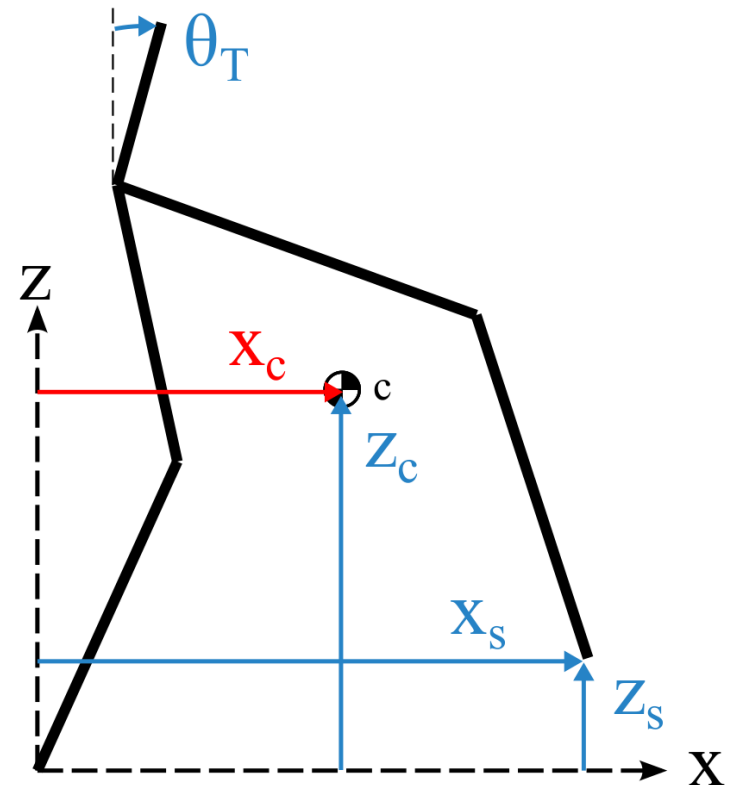
$$L_y = mz_c\dot{x}_c - mx_c\dot{z}_c + H_c$$

Take the derivative and substitute for $\dot{L}_y = mgx_c$

$$mgx_c = m\dot{z}_c\dot{x}_c + mz_c\ddot{x}_c - m\dot{x}_c\dot{z}_c - mx_c\ddot{z}_c + \dot{H}_c$$

Assuming constant center of mass height and low \dot{H}_c

$$mgx_c = mz_c\ddot{x}_c$$



Returning to basics!

Harold Black



Amplifier



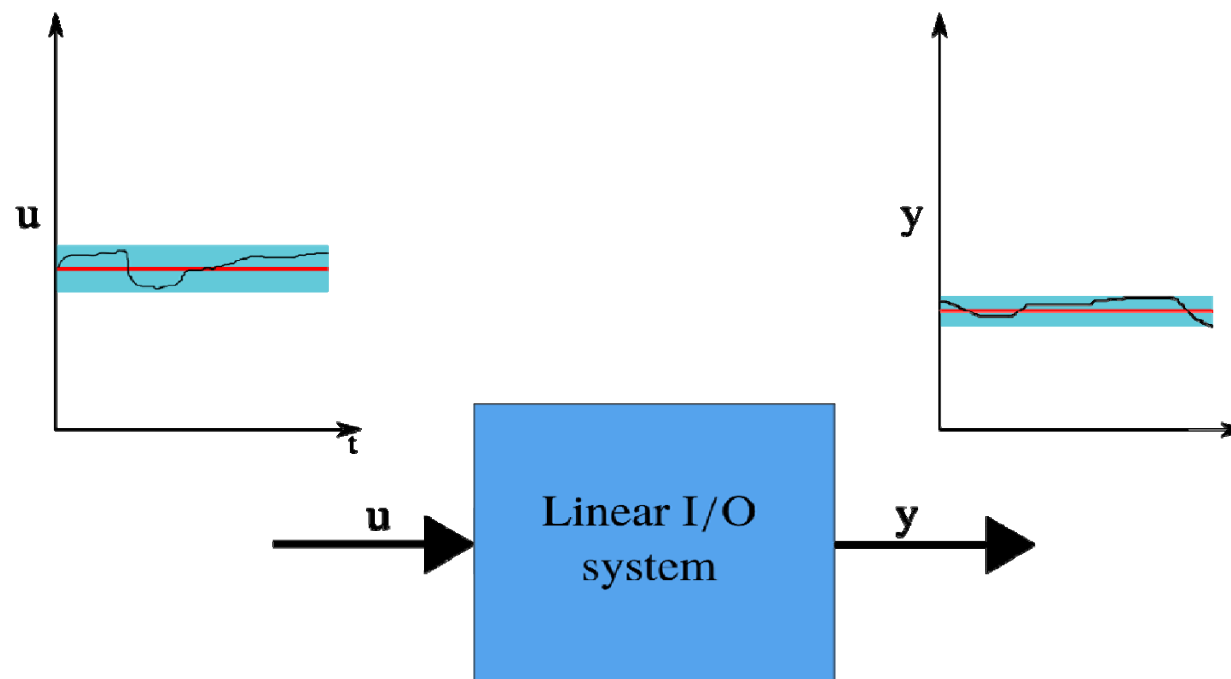
George Zames



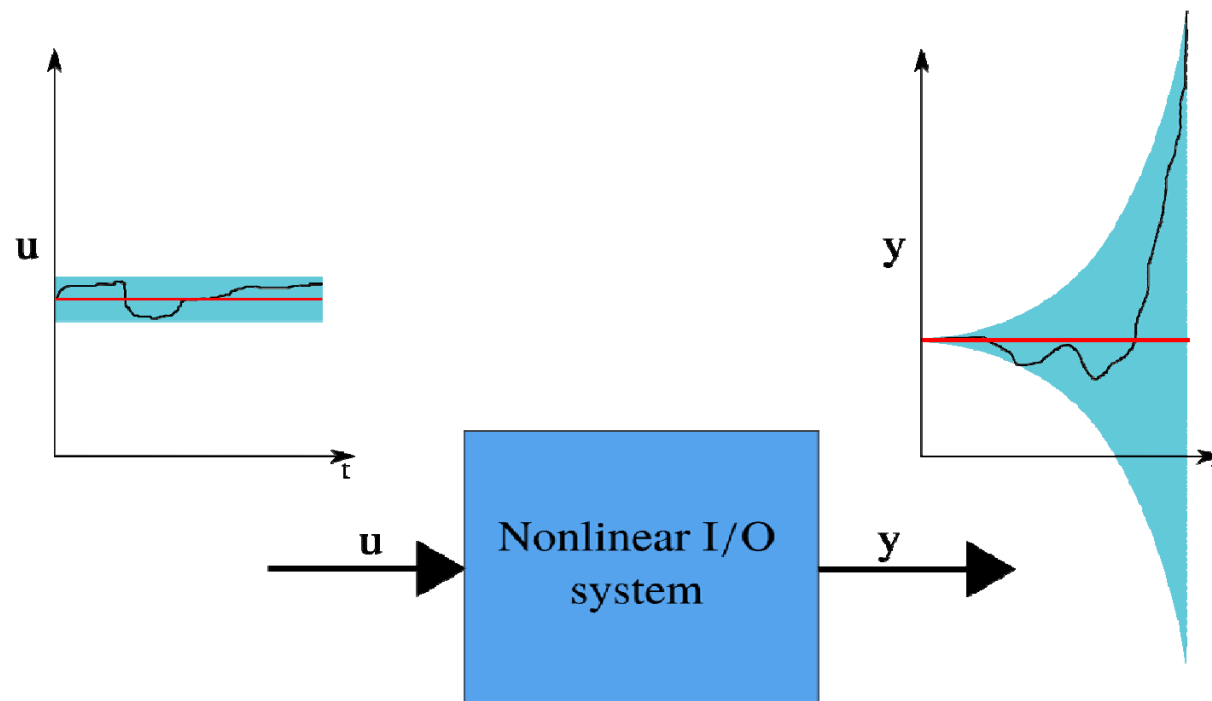
Black: Negative feedback for amplifiers.

Zames: Input/output stability. Operator theory.

Returning to basics!



Preliminaries on ISS



Preliminaries on ISS

George Zames



Input/output stability
for linear systems

Aleksandr Lyapunov



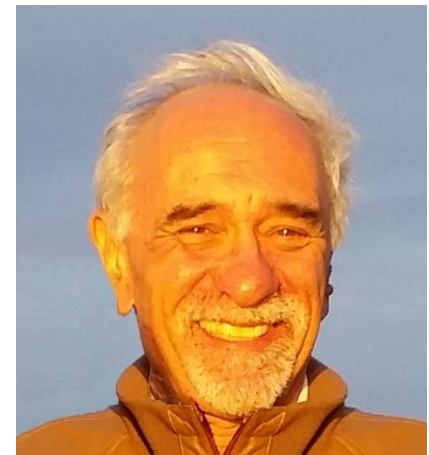
Lyapunov stability for
nonlinear systems

+

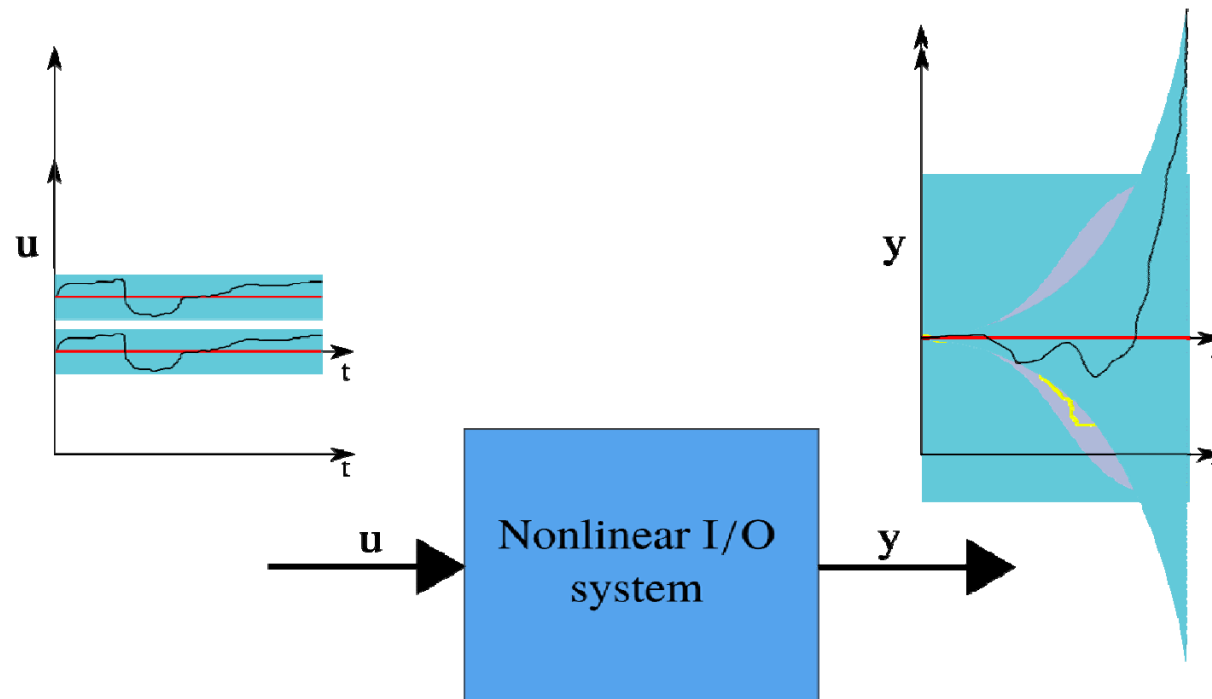
=

Input to state stability
for nonlinear systems

Eduardo Sontag



Preliminaries on ISS



Properties of ISS

ISS is always w.r.t. d .

$$\dot{x} = f(x, k(x) + d), \quad u = k(x) + d, \quad x(0) = x_0$$

If $\dot{x} = f(x, k(x))$ is asymptotically stable, then $u = k(x)$ is the stabilizing controller.

If $\dot{x} = f(x, k(x) + d)$ is input to state stable, then $u = k(x)$ is the input to state stabilizing controller.

How to obtain ISSing controllers?

Linear feedback laws

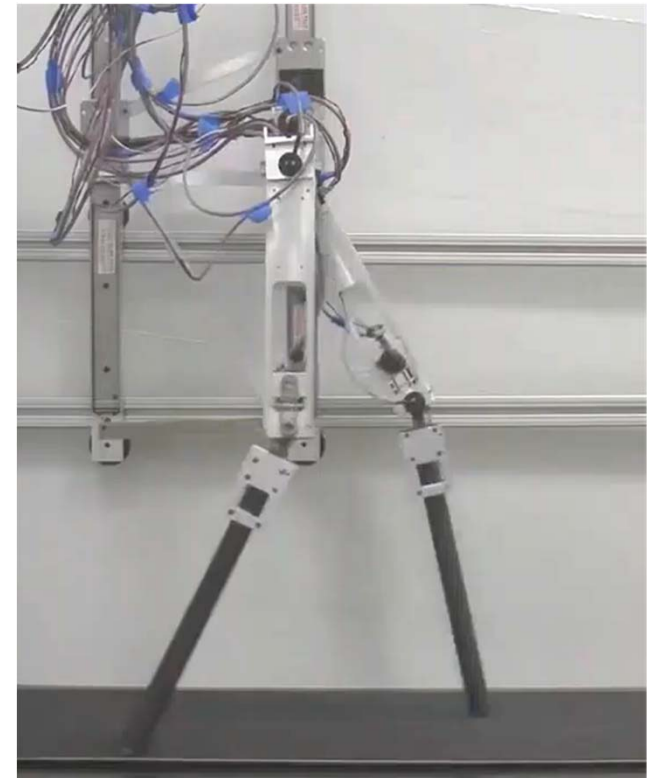
$$D(q)\ddot{q} + \boxed{C(q, \dot{q}) + d} G(q) = BK_p(q_d - q) + BK_d(\dot{q}_d - \dot{q})$$

Stabilizing controller

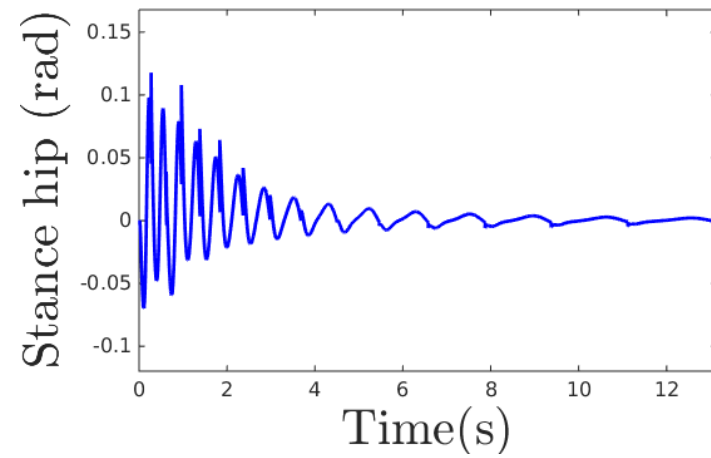
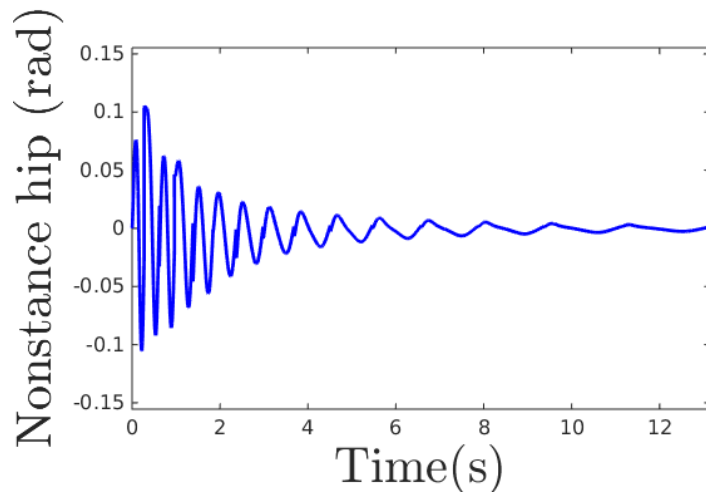
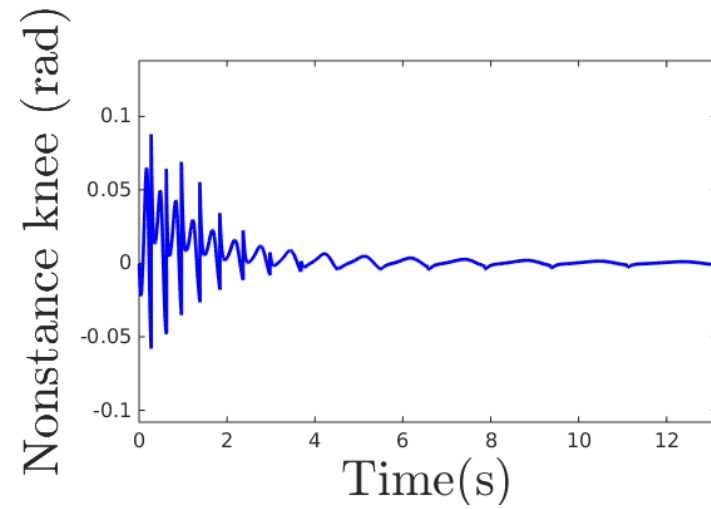
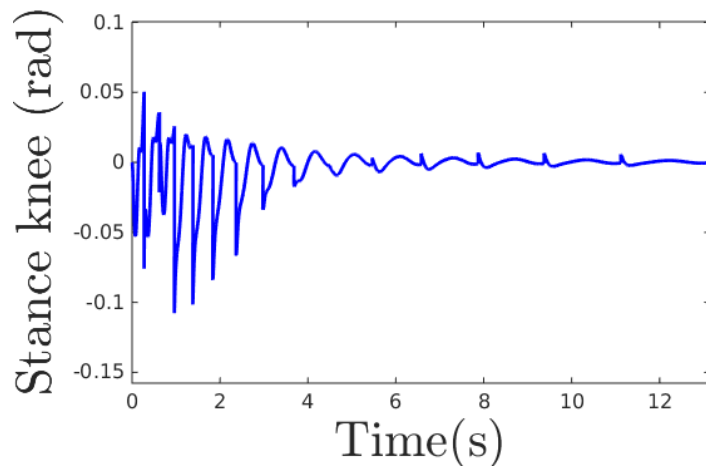
+

Hybrid invariance

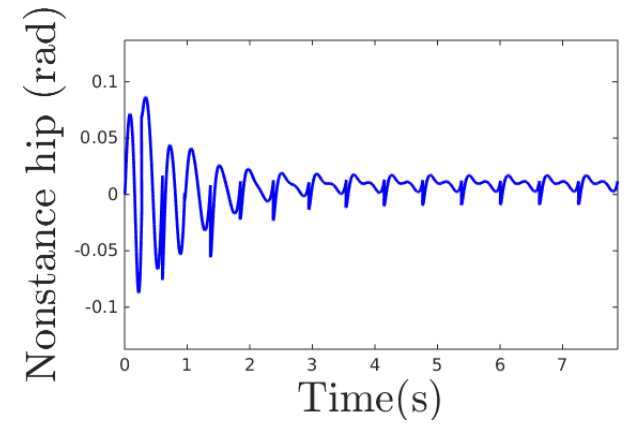
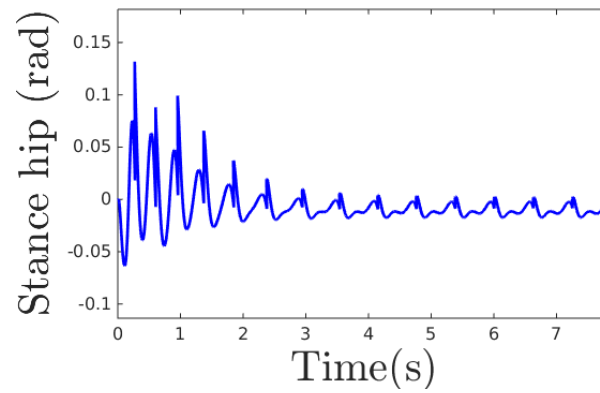
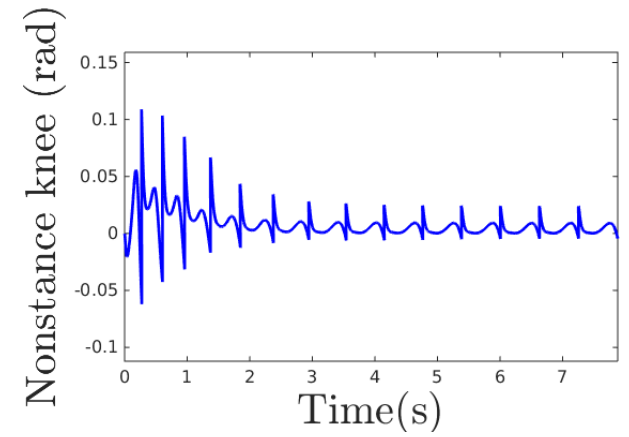
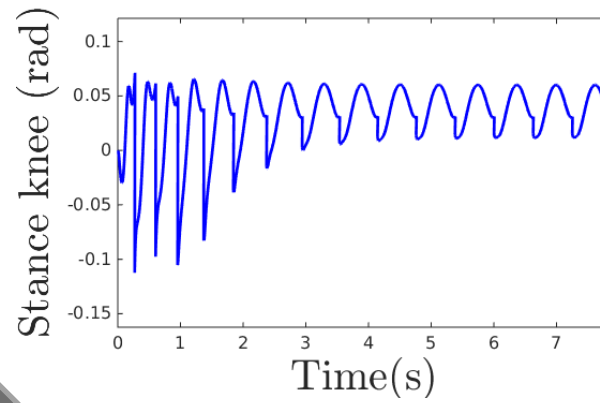
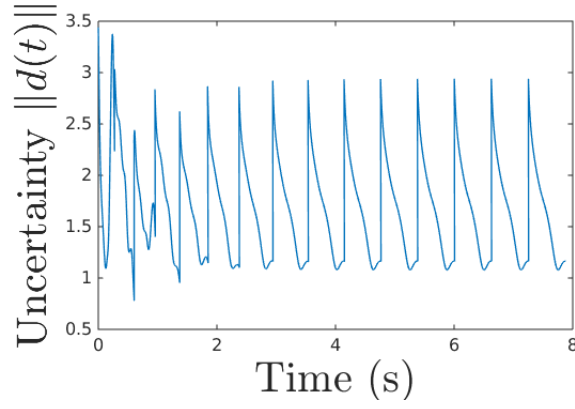
= Output stability



$$d = 0$$



$$d \neq 0$$

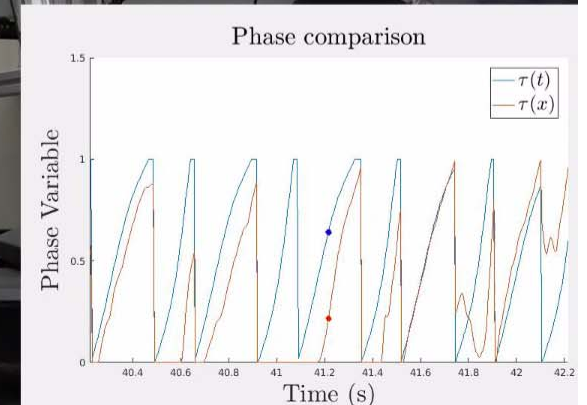
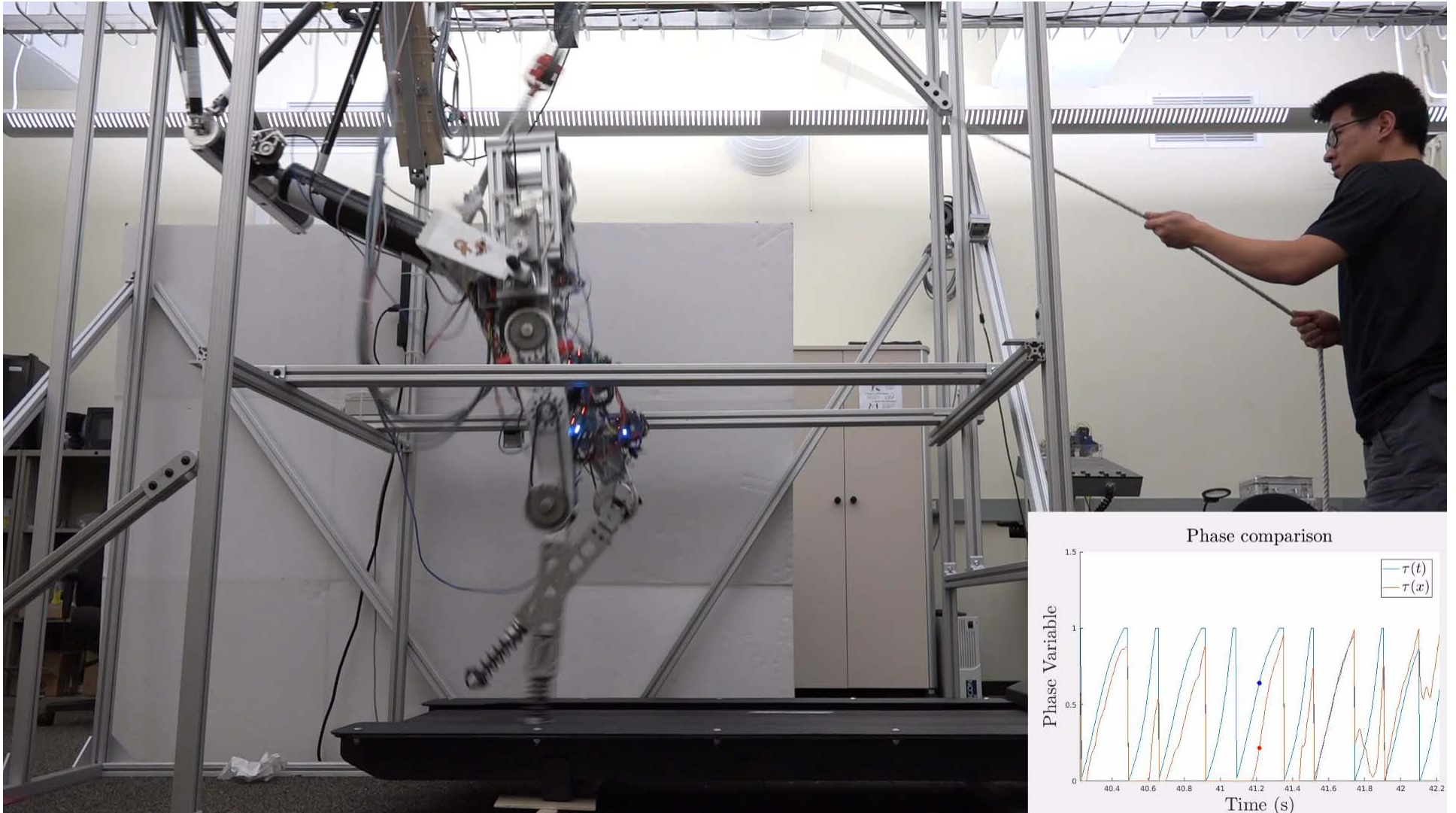


Phase uncertainty

$$y = y_a - y_d = \begin{bmatrix} y_{sk} \\ y_{sh} \\ y_{nsh} \\ y_{nsk} \end{bmatrix} - y_d(\tau(q))$$



DURUS-2D Running with ISS based controllers



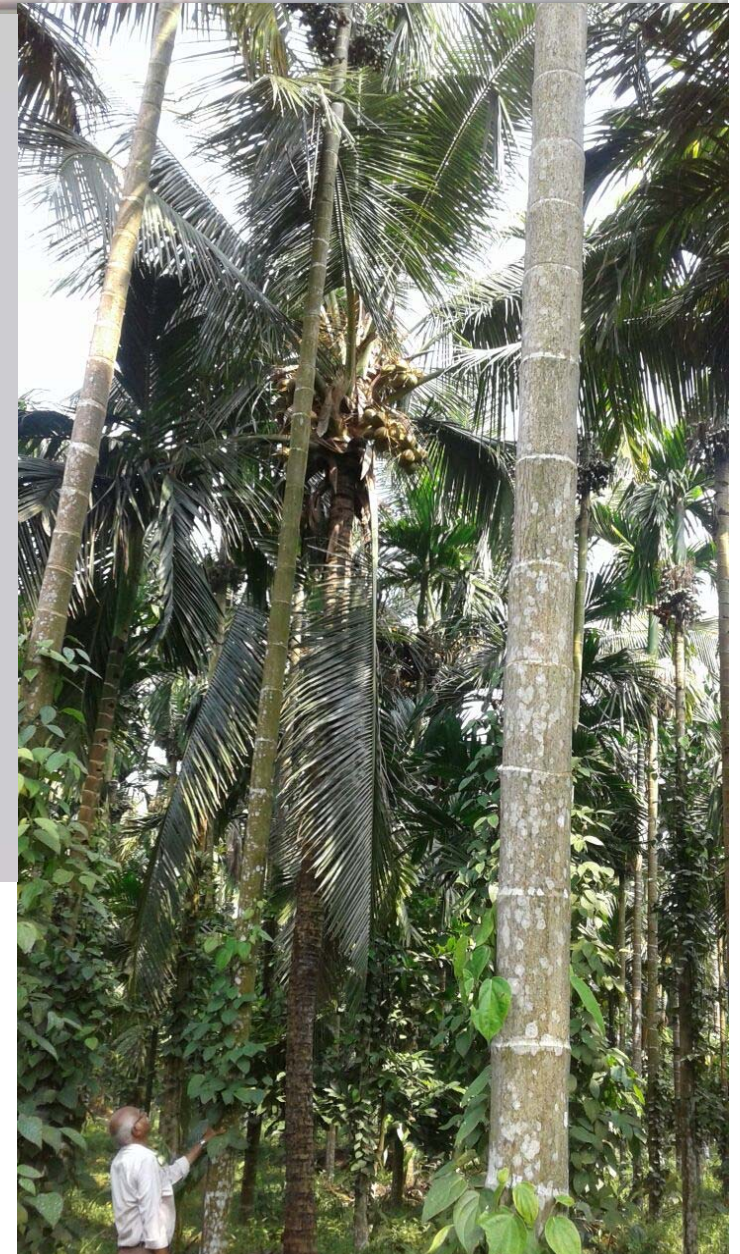
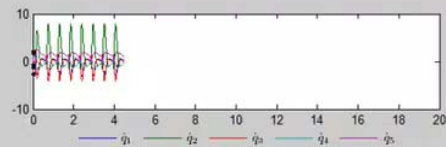
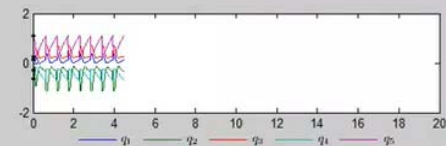
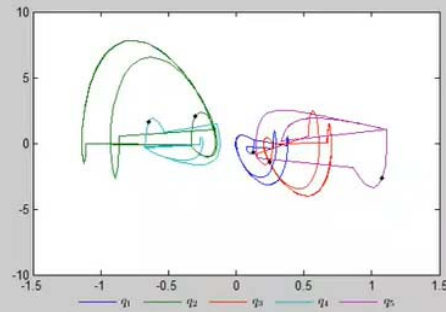
Conclusions

Made a significant step toward bridging the gap between theory and experiment

- Gait design: Demonstrated fast gait generation via collocation methods.
- Controller design: Constructed ISSing controllers to address modeling and phase based uncertainties.



Next steps: Food and agriculture



Challenges

- Hybrid system model
- Underactuation
- Soft contact with trunk



Robots in agriculture



Farming and mining applications will be the first to receive self-driving tech in India, because the driving patterns are far simpler – no humans and other vehicles in your path to worry about. They're also quick to adopt new technologies. Plus, these industries are key drivers of our country's economy.

-- Dr. Roshy John

Thank you!



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