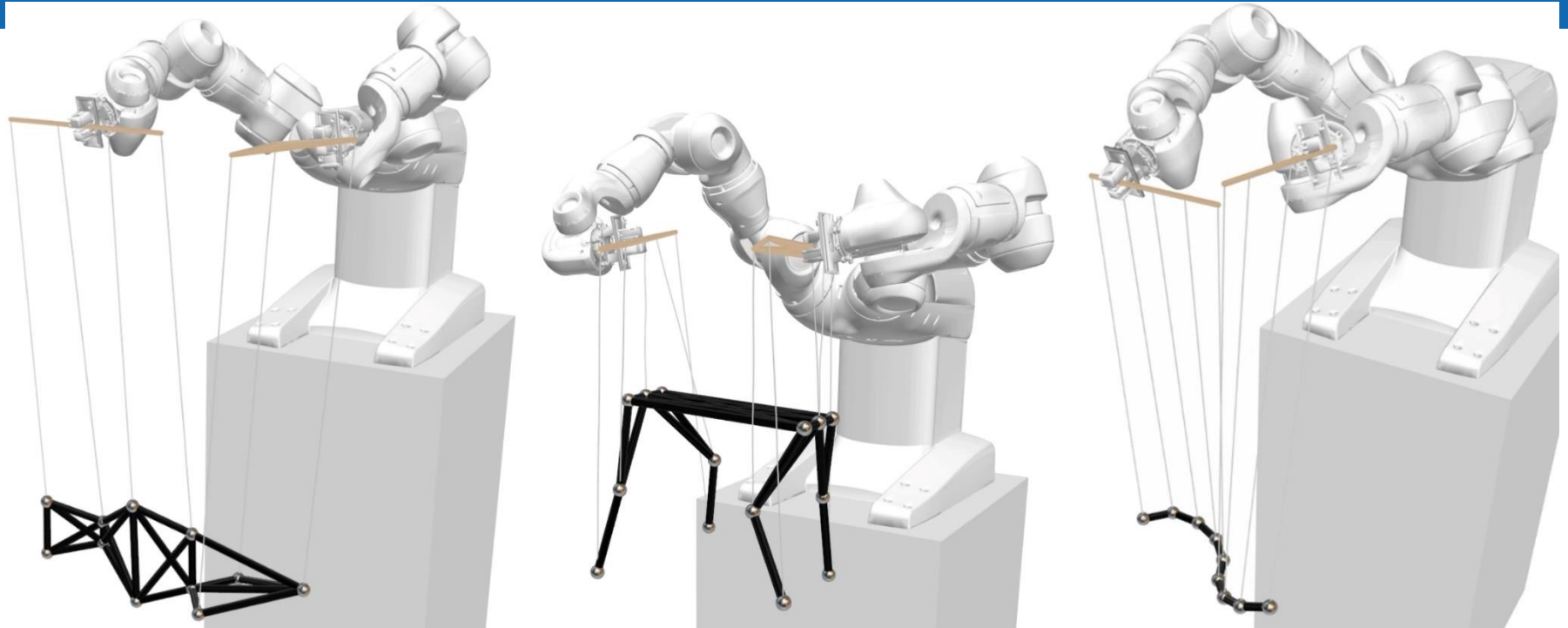




Puppet Master

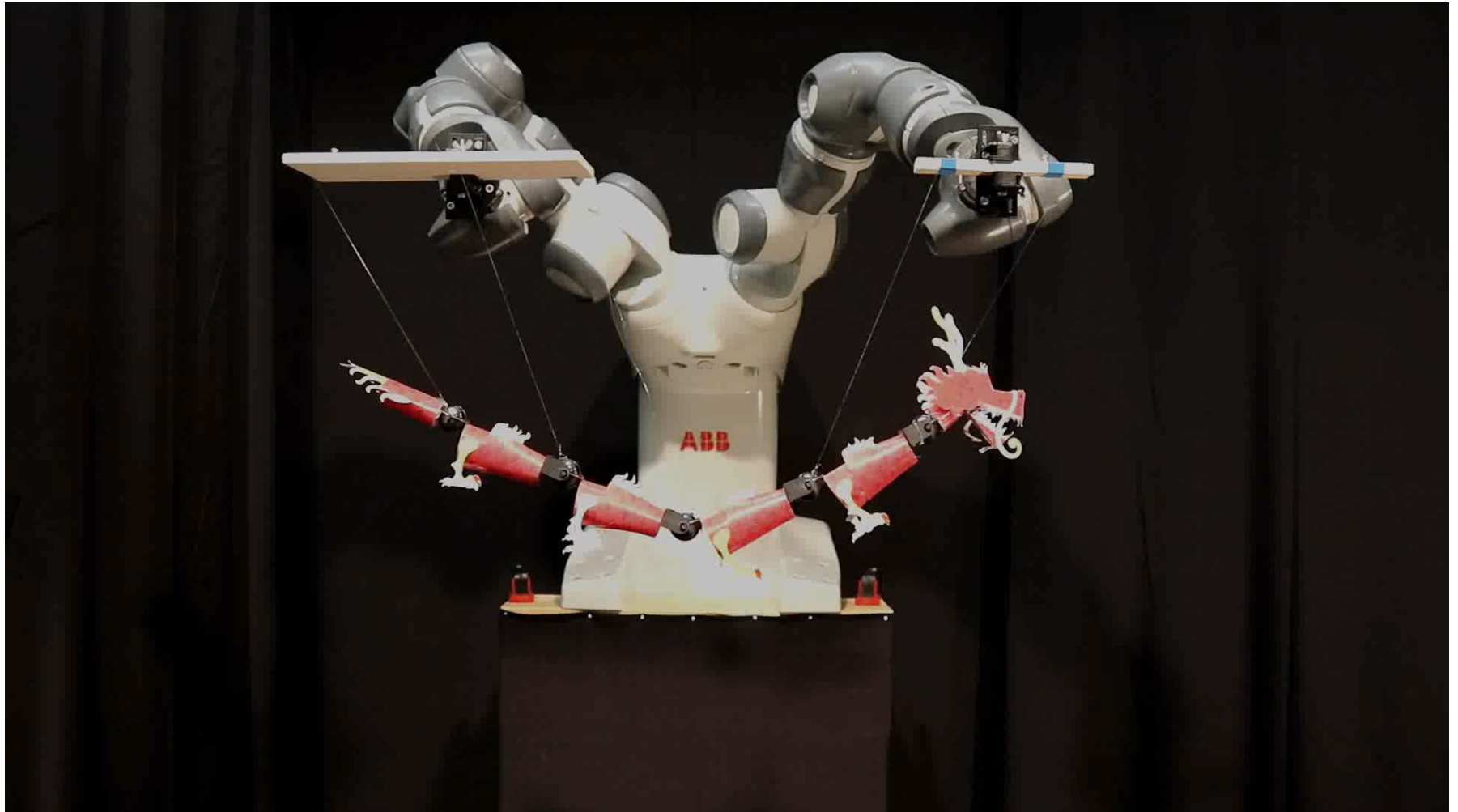
Simon Zimmermann

Dr. Roi Poranne

James Bern

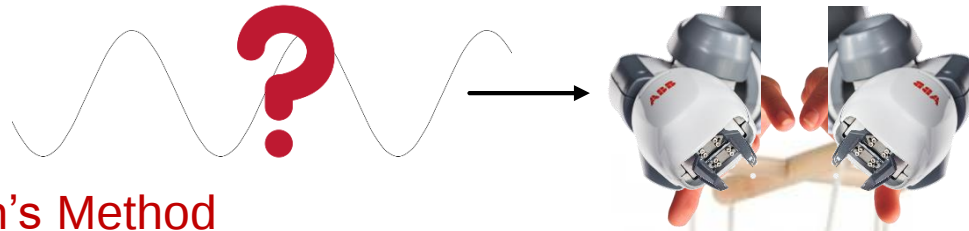
Prof. Dr. Stelian Coros

First Example



Introduction

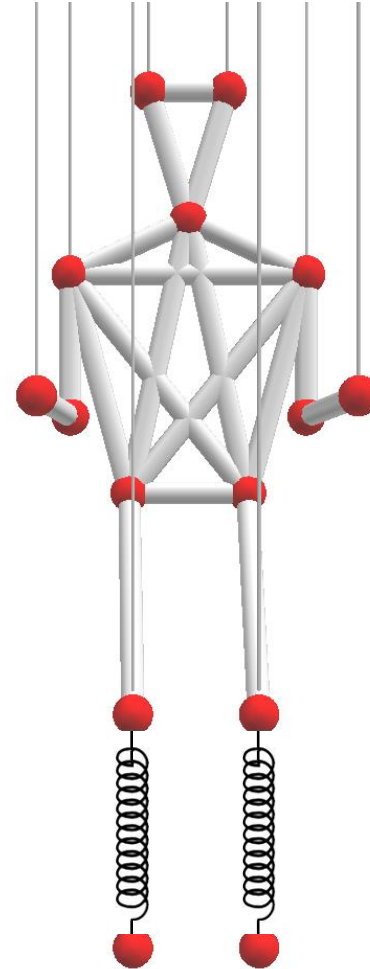
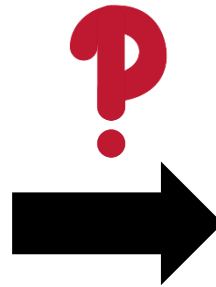
Newton's Method



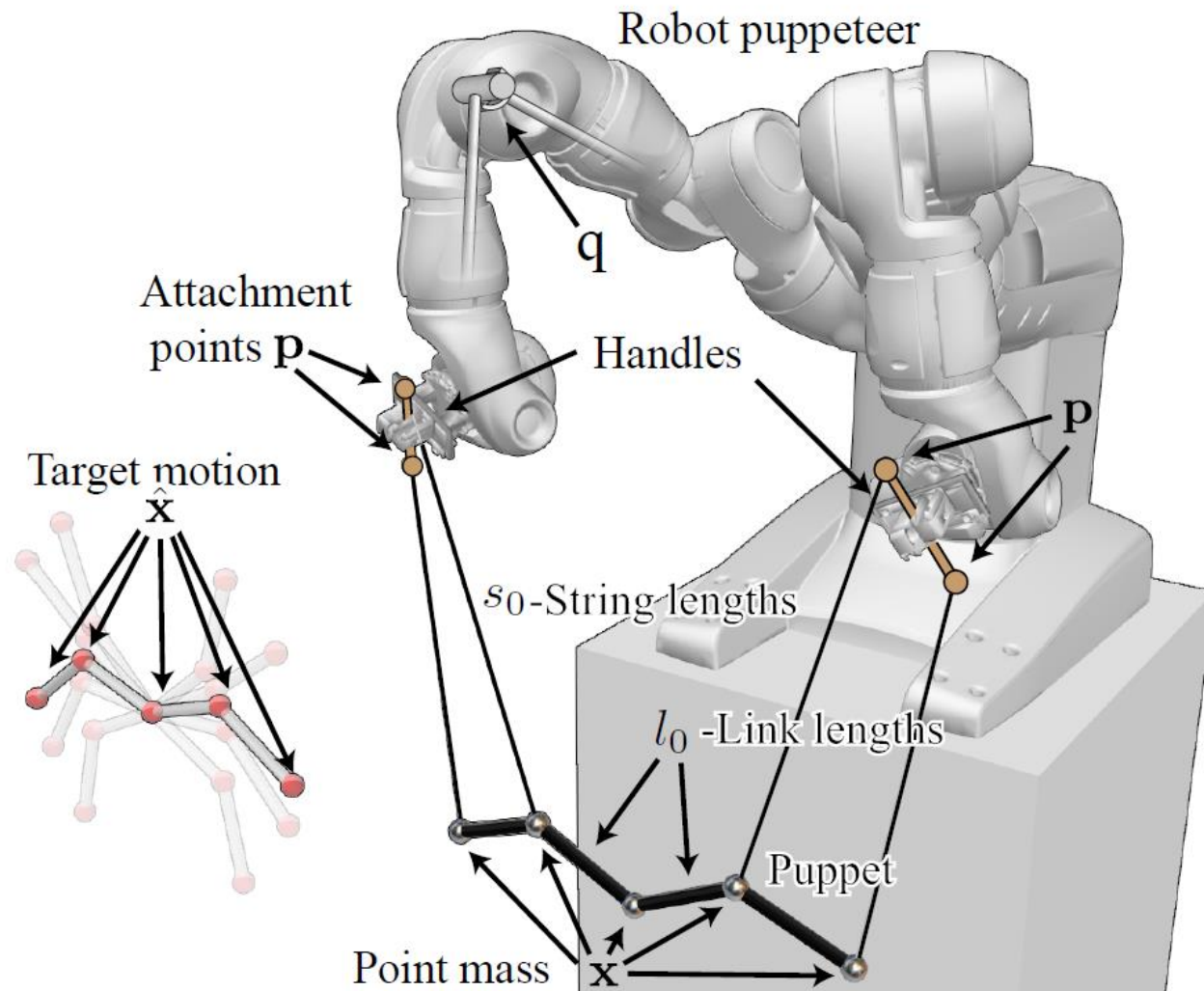
Forward Simulation



Modelling



System Overview



Dynamics – Forward Integration

$$\dot{x} = f(t, x), \quad x(t_0) = x_0$$

Implicit:

$$x_{k+1} = x_k + hf(t_{k+1}, x_{k+1})$$

Explicit:

$$x_{k+1} = x_k + hf(t_k, x_k)$$

Dynamics – Forward Simulation

$$\mathbf{x}_i(\mathbf{p}_i) = \underset{\mathbf{x}_i}{\operatorname{argmin}} U_i(\mathbf{x}_i, \mathbf{p}_i)$$

where
$$U_i(\mathbf{x}_i, \mathbf{p}_i) = \frac{h^2}{2} \ddot{\mathbf{x}}_i^T M \ddot{\mathbf{x}}_i + W(\mathbf{x}_i, \mathbf{p}_i) + \mathbf{g}^T M \mathbf{x}_i$$

$$\longrightarrow \ddot{\mathbf{x}}_i \approx \frac{\mathbf{x}_i - 2\mathbf{x}_{i-1} + \mathbf{x}_{i-2}}{h^2}$$

$$\longrightarrow W(\mathbf{x}_i, \mathbf{p}_i) = W_{\text{puppet}}(\mathbf{x}_i) + W_{\text{string}}(\mathbf{x}_i, \mathbf{p}_i)$$



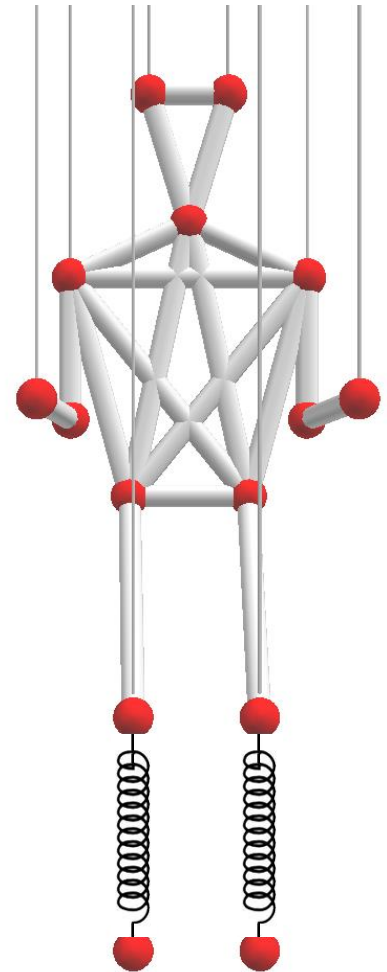
Dynamics – Forward Simulation

$$W(\mathbf{x}_i, \mathbf{p}_i) = W_{puppet}(\mathbf{x}_i) + W_{string}(\mathbf{x}_i, \mathbf{p}_i)$$

$$W_{puppet}(\mathbf{x}_i) = \sum_{(j_1, j_2) \in L} k_s \left(\|\mathbf{x}_i^{j_1} - \mathbf{x}_i^{j_2}\| - l_0^{j_1, j_2} \right)^2$$

$$W_{string}(\mathbf{x}_i) = \sum_{(j_1, j_2) \in S} k_s \psi \left(\|\mathbf{x}_i^{j_1} - \mathbf{x}_i^{j_2}\| - s_0^{j_1, j_2} \right)$$

$$\psi(x) = \begin{cases} \frac{1}{2}x^2 + \frac{\epsilon}{2}x + \frac{\epsilon^2}{6} & x > 0 \\ \frac{1}{6\epsilon}x^3 + \frac{1}{2}x^2 + \frac{\epsilon}{2}x + \frac{\epsilon^2}{6} & x > -\epsilon \\ 0 & \text{otherwise} \end{cases}$$



Control – Sensitivity Analysis

Objective: $\min_p O(x(p), p) \longrightarrow O(x(p), p) = \|x - \hat{x}\|_2^2 + \|\ddot{p}\|_2^2$

Gradient: $\frac{dO}{dp} = \frac{\partial O}{\partial x} \frac{dx}{dp} + \frac{\partial O}{\partial p}$

➡ $G(x, p) = 0 \longrightarrow \frac{dG}{dp} = 0 = \frac{\partial G}{\partial x} \frac{dx}{dp} + \frac{\partial G}{\partial p}$

$\longrightarrow \frac{dx}{dp} = - \left(\frac{\partial G}{\partial x} \right)^{-1} \frac{\partial G}{\partial p} := S$

$f - ma = 0$

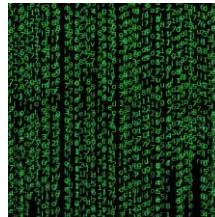
➡ $\nabla_{x_i} U_i(x_i) = \nabla_{x_i} \left(\frac{h^2}{2} \ddot{x}_i^T M \ddot{x}_i + W(x_i, p_i) + g^T M x_i \right)$

$= M \ddot{x}_i + \nabla_{x_i} W(x_i, p_i) + M g = M \ddot{x}_i - F(x_i, p_i) := G$

Second Order Sensitivity Analysis

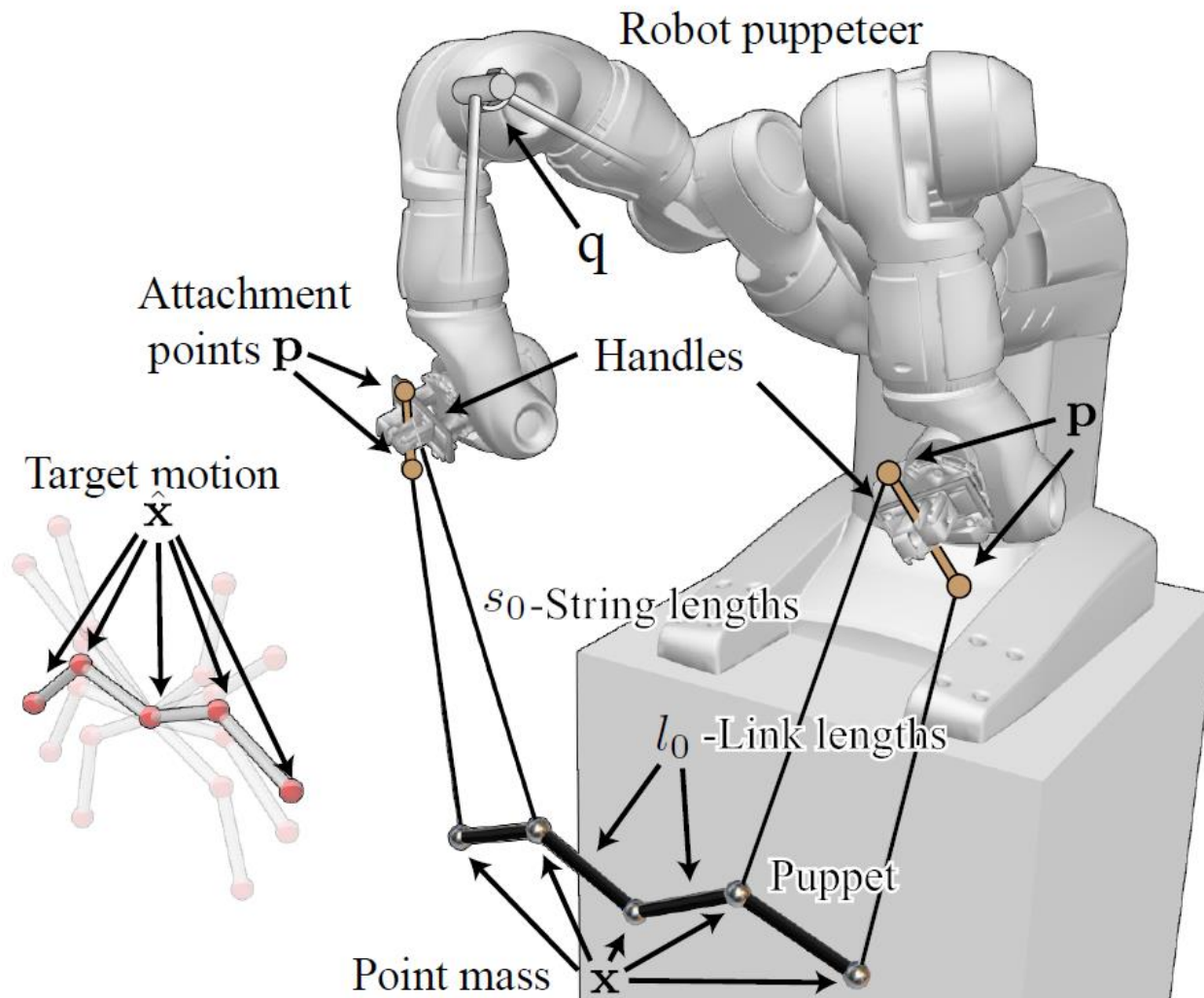
Newton's Method: $\frac{d^2 O}{d\mathbf{p}^2} \Delta \mathbf{p} = -\frac{dO}{d\mathbf{p}}$

$$\frac{d^2 O}{d\mathbf{p}^2} = \frac{d}{d\mathbf{p}} \frac{dO}{d\mathbf{p}} = \frac{d}{d\mathbf{p}} \left(\frac{\partial O}{\partial \mathbf{x}} \mathbf{s} + \frac{\partial O}{\partial \mathbf{p}} \right)$$



$$\frac{d^2 O}{d\mathbf{p}^2} = \frac{\partial O}{\partial \mathbf{x}} \left(\cancel{\mathbf{s}^T \frac{\partial}{\partial \mathbf{x}} \mathbf{s} + \frac{\partial}{\partial \mathbf{p}} \mathbf{s}} \right) + \mathbf{s}^T \left(\frac{\partial^2 O}{\partial \mathbf{x}^2} \mathbf{s} + 2 \frac{\partial O}{\partial \mathbf{x} \partial \mathbf{p}} \right) + \frac{\partial^2 O}{\partial \mathbf{p}^2}$$

Joint Space

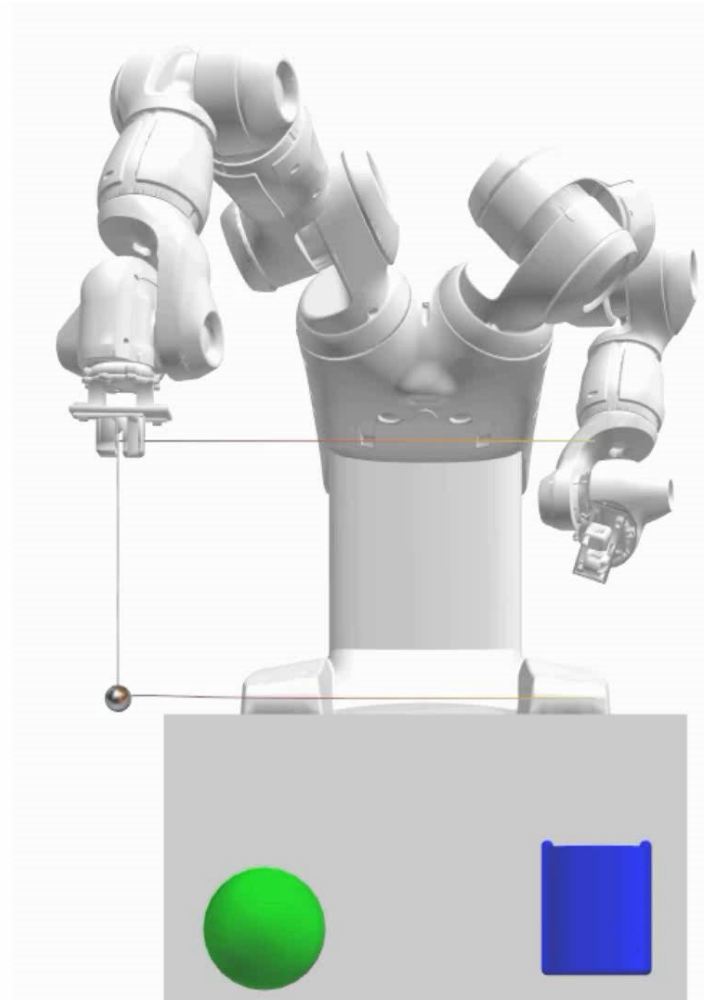


$$\min_q O(x(p(q)), p, q)$$

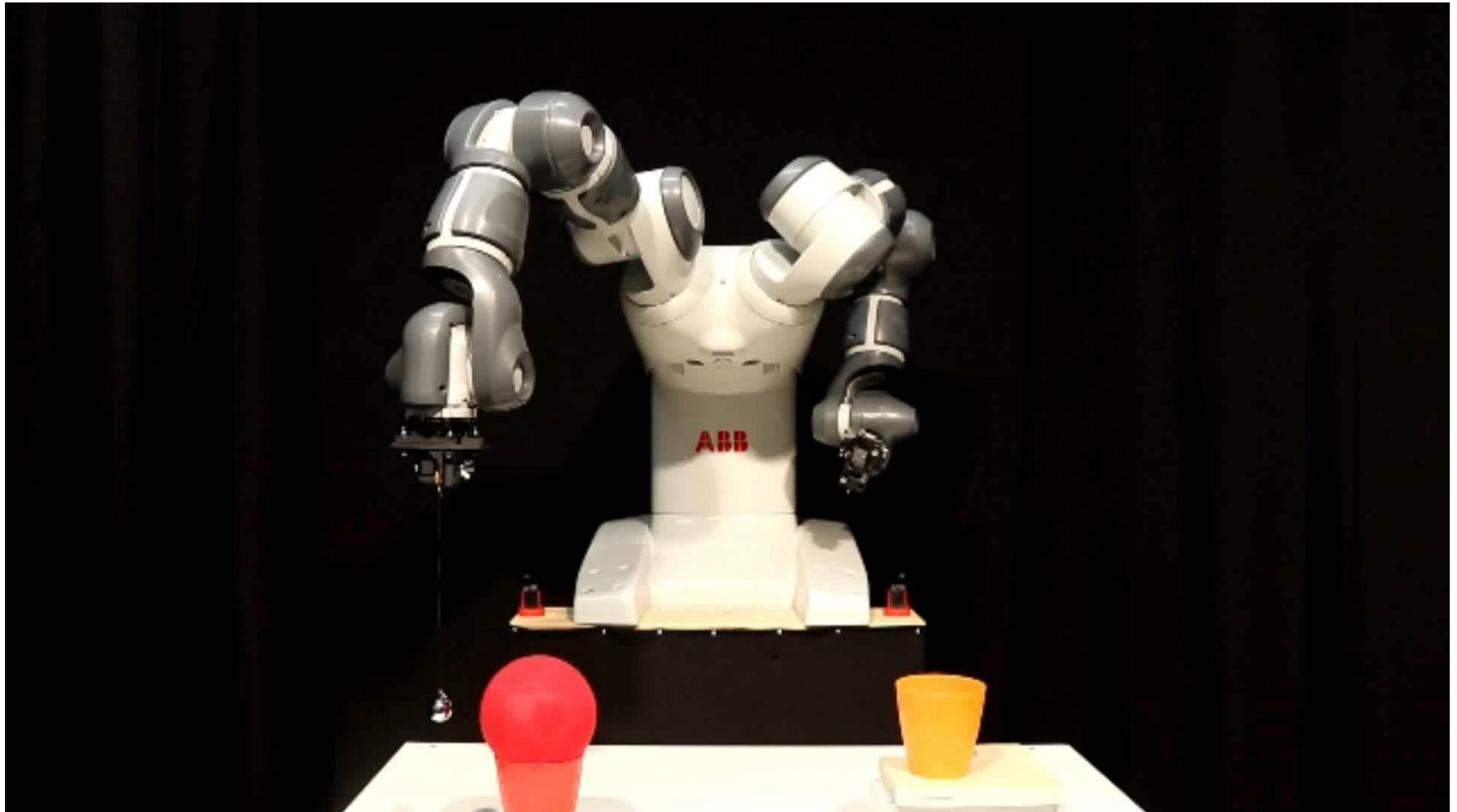
$$\frac{dO}{dq} = \frac{\partial p^T}{\partial q} \frac{dO}{dp} + \frac{\partial O}{\partial q}$$

Preliminary Experiments – Pendulum

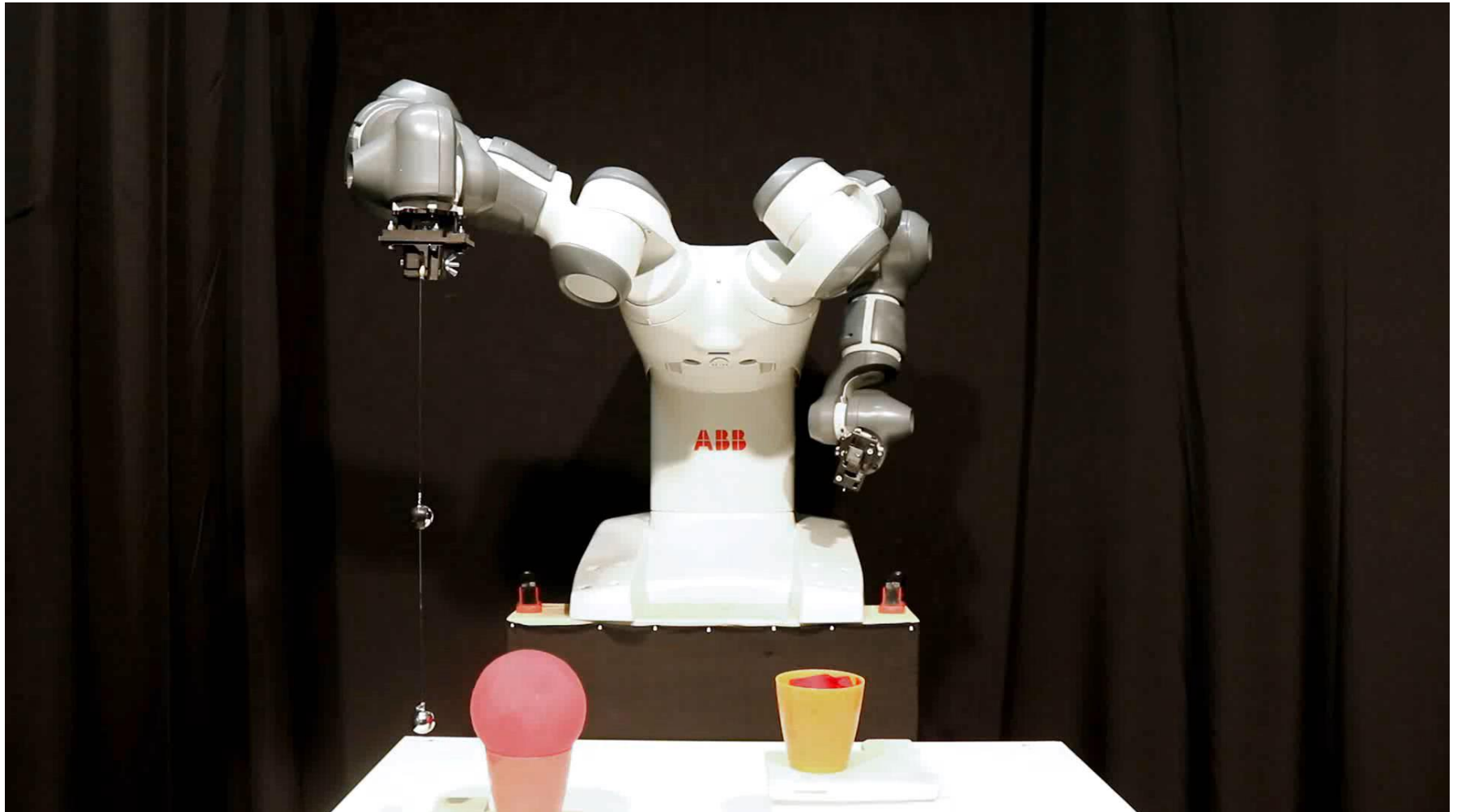
Task:
Avoid obstacle
and jump into cup
(robot hand constrained
to move along horizontal
axis)



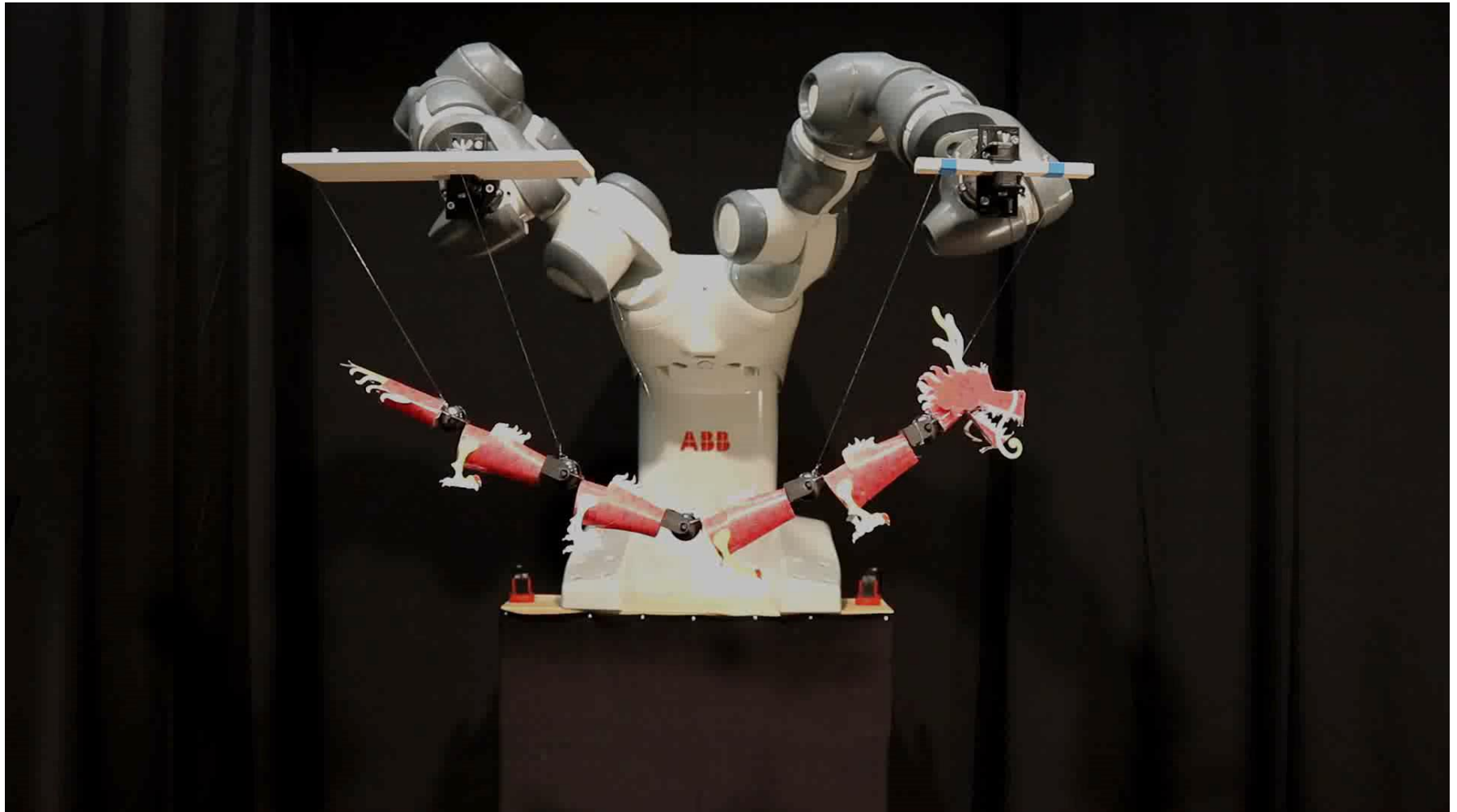
Preliminary Experiments – Pendulum



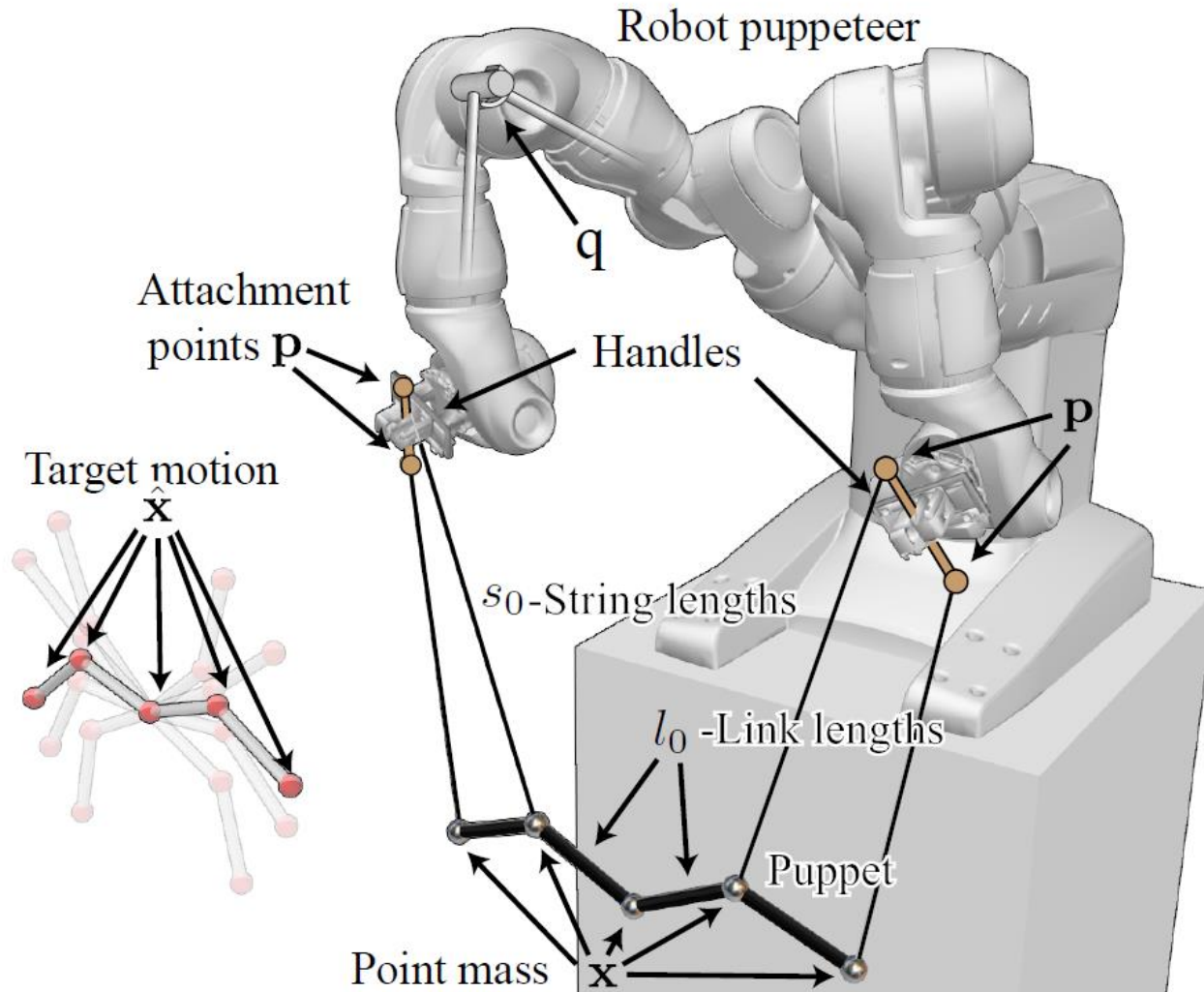
Preliminary Experiments – Double Pendulum



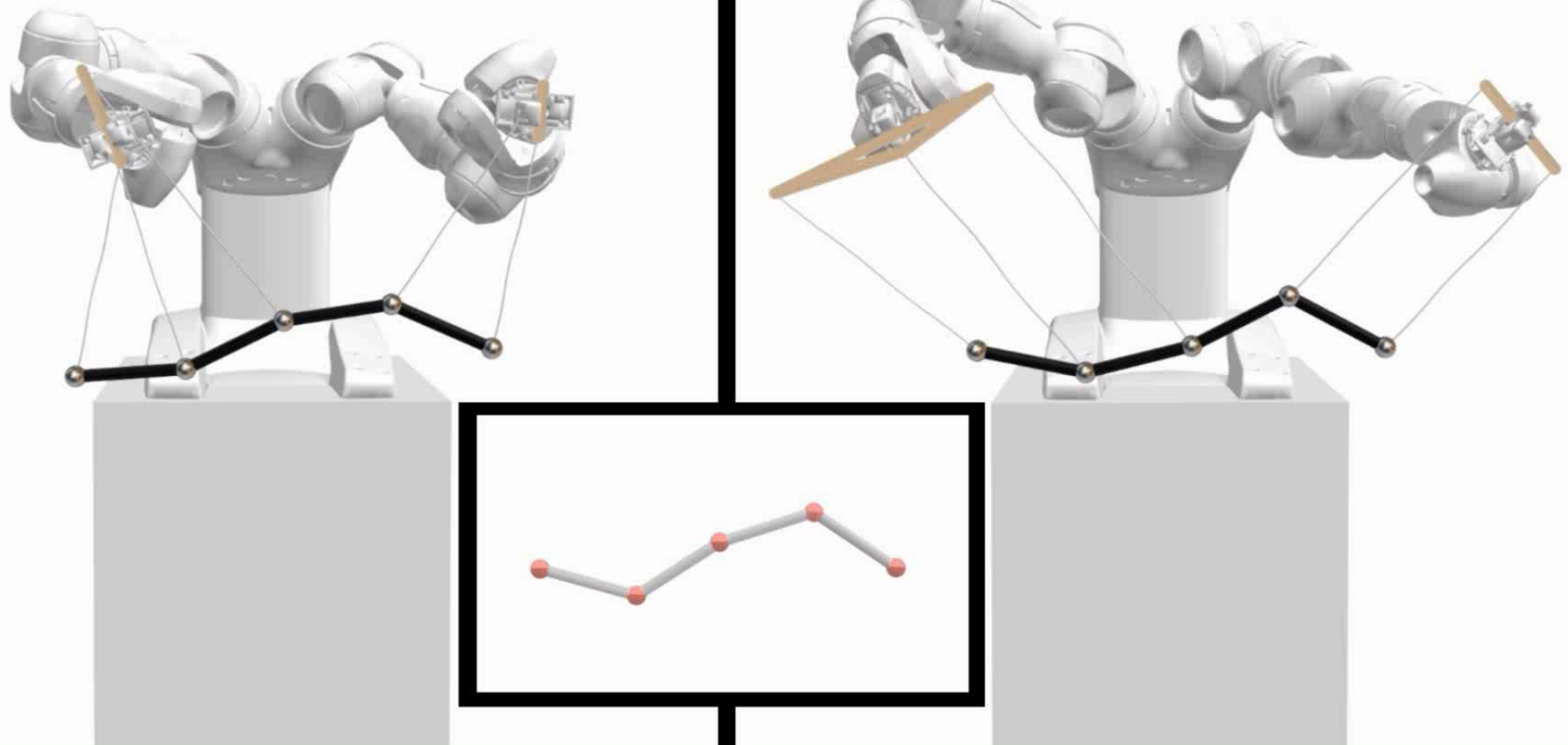
Physical Puppet: Chinese Dragon



Design Optimization



Design Optimization



Simulation Results

Simulation Results

Physical Results

Physical Results

