

Feedback for Stabilization & Control of a Two-Link Arm

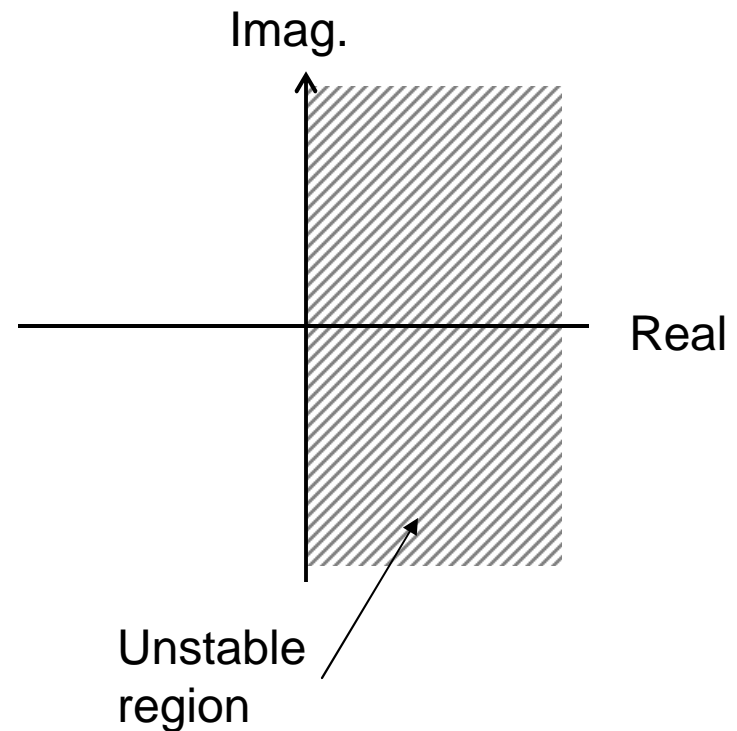
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Stabilization via Feedback

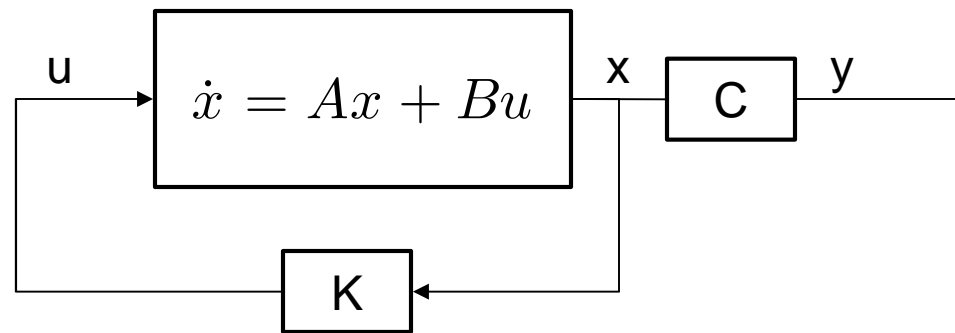
- Given an autonomous system (no external control), we have learned how to determine instability:
 - Poles lie in the right half plane, or, equivalently:
 - Eigenvalues of A lie in the right half plane
- If A is unstable, we can try to stabilize this system via feedback
 - State feedback
 - Output feedback

$$\dot{x} = Ax$$



State Feedback Design

- If states are directly measurable, can create the following system:



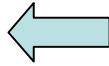
- Substituting $u=Kx$, this system is equivalent to:

$$\dot{x} = \underbrace{(A + BK)}_M x$$

- If we choose K such that the real parts of the eigenvalues of M are negative, we have stabilized the system. (This is not always possible.)

State Feedback Example

$$\begin{aligned}\dot{x} &= Ax + Bu \\ u &= -Kx\end{aligned}$$



This A is unstable (just trust us!)

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 2 & -1 & 4 & 3 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$K = [k_1 \quad k_2 \quad k_3 \quad k_4]$$

$$\dot{x} = (A + BK)u = Mx$$

$$M = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 2 & -1 & 4 & 3 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} [k_1 \quad k_2 \quad k_3 \quad k_4] = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 2 - k_1 & -1 - k_2 & 4 - k_3 & 3 - k_4 \end{bmatrix}$$

We use the characteristic polynomial to find the eigenvalues of M:

$$\det(\lambda I - M) = \lambda^4 - (3 - k_4)\lambda^3 - (4 - k_3)\lambda^2 - (-1 - k_2)\lambda - (2 - k_1) = 0$$

One possible stable M would have all eigenvalues = -1. That would yield:

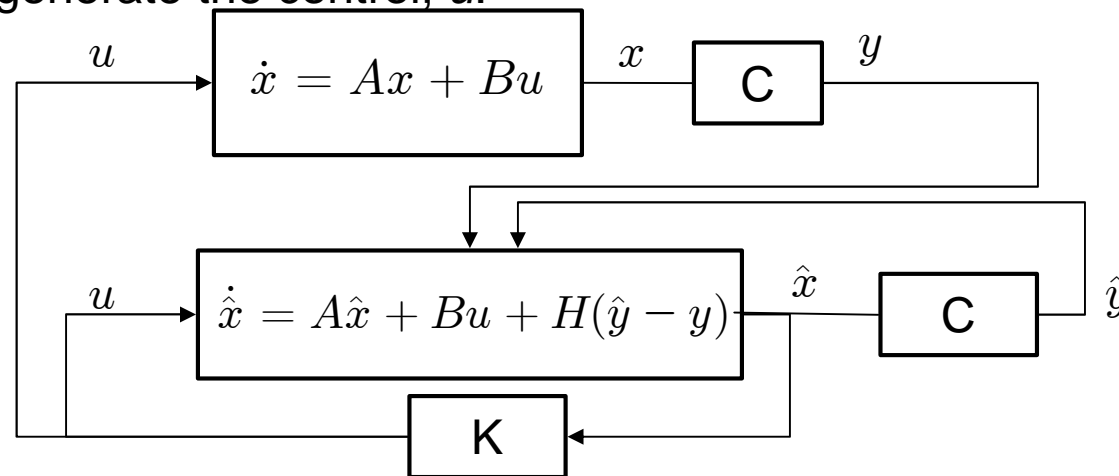
$$(\lambda + 1)^4 = \lambda^4 + 4\lambda^3 + 6\lambda^2 + 4\lambda + 1 = 0$$

Matching coefficients of ? and solving for the k_i :

$$K = [3 \quad 3 \quad 10 \quad 7]$$

Output Feedback Design

- Usually direct state measurements are not available. Then we must use the output as feedback.
- This involves using a *closed-loop observer*, which simulates the actual system to generate the control, u :



- The resulting system has the following state dynamics:

$$\begin{bmatrix} \dot{x} \\ \dot{\hat{x}} \end{bmatrix} = \underbrace{\begin{bmatrix} A & BK \\ -HC & A + BK + HC \end{bmatrix}}_M \begin{bmatrix} x \\ \hat{x} \end{bmatrix}$$

- We need to choose K and H such that M will be stable. (Not always possible.)
- There is a separation principle that allows us to find K and H separately for the matrices $A+BK$ and $A+HC$. (It decouples the problem.)

Control Example: Two-link Robot Arm

- Kinetic Energy:

$$v_i = \begin{bmatrix} \dot{x}_i \\ \dot{y}_i \end{bmatrix} = \begin{bmatrix} -l_i \sin(q_i) \dot{q}_i \\ l_i \cos(q_i) \dot{q}_i \end{bmatrix}$$

$$K = \frac{1}{2}m_1 \|v_1\|^2 + \frac{1}{2}m_2 \|v_2\|^2$$

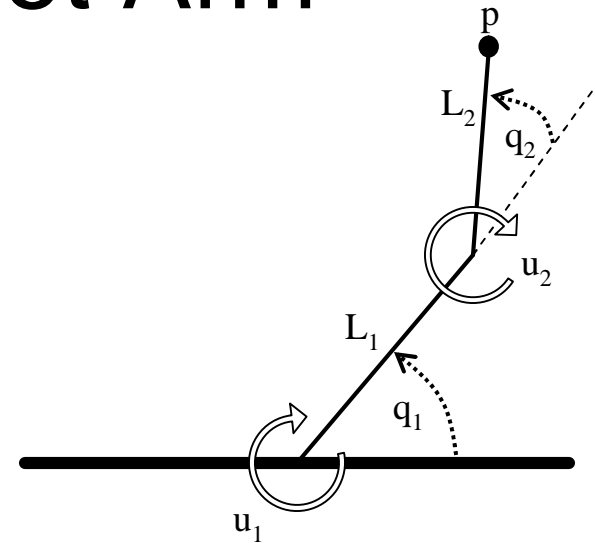
- Potential Energy:

$$P(q) = m_1 g l_1 \sin(q_1) + m_2 g (l_2 \sin(q_1) + l_2 \sin(q_1 + q_2))$$

- Lagrangian:

$$L = K - P$$

$$\frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}} \right) - \frac{\partial L}{\partial q} = u$$



Notation:

- q_i = Joint Angles
- \dot{q}_i = Joint [angular] velocities
- p = end effector coordinates
- L_i = Link lengths
- m_i = Link masses
- u_i = Joint torques (controls)
- g = gravity

Deriving Arm Dynamics

- $D(q)$ is the *mass matrix* defined (via ugly trigonometry) such that:

$$K = \frac{1}{2} \dot{q}^T D(q) \dot{q}$$

- Using the Lagrangian, we can obtain the dynamics equation:

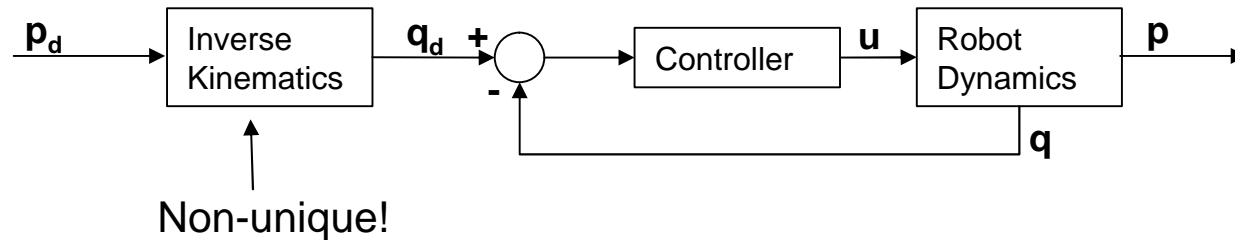
$$\begin{array}{ccc} & \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}} \right) - \frac{\partial L}{\partial q} = u & \\ \nearrow & & \nwarrow \\ \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{q}} \right) = D(q) \ddot{q} + \dot{D}(q) \dot{q} & & \frac{\partial L}{\partial q} = \frac{\partial K}{\partial q} - \frac{\partial P}{\partial q} \\ \uparrow & & \\ \frac{\partial L}{\partial \dot{q}} = D(q) \dot{q} & & \end{array}$$

$$\underbrace{D(q) \ddot{q}} + \underbrace{\dot{D}(q) \dot{q}} - \underbrace{\frac{\partial K}{\partial q}} + \underbrace{\frac{\partial P}{\partial q}} = u$$

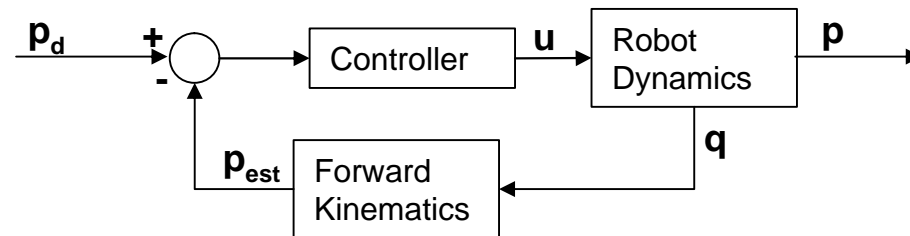
- Final form: $\boxed{D(q) \ddot{q} + V(q, \dot{q}) + G(q) = u}$

Control Strategies

- Joint Space Control

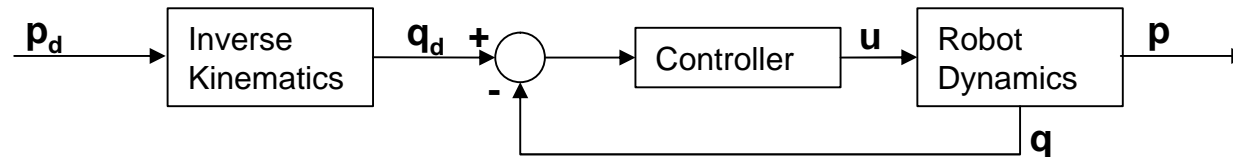


- Task Space Control



- In general, forward kinematics are “easier,” but possible singularities can occur in the Task Space Controller box. So there is an inherent tradeoff between the two approaches.

Classical [Linear] Joint Control

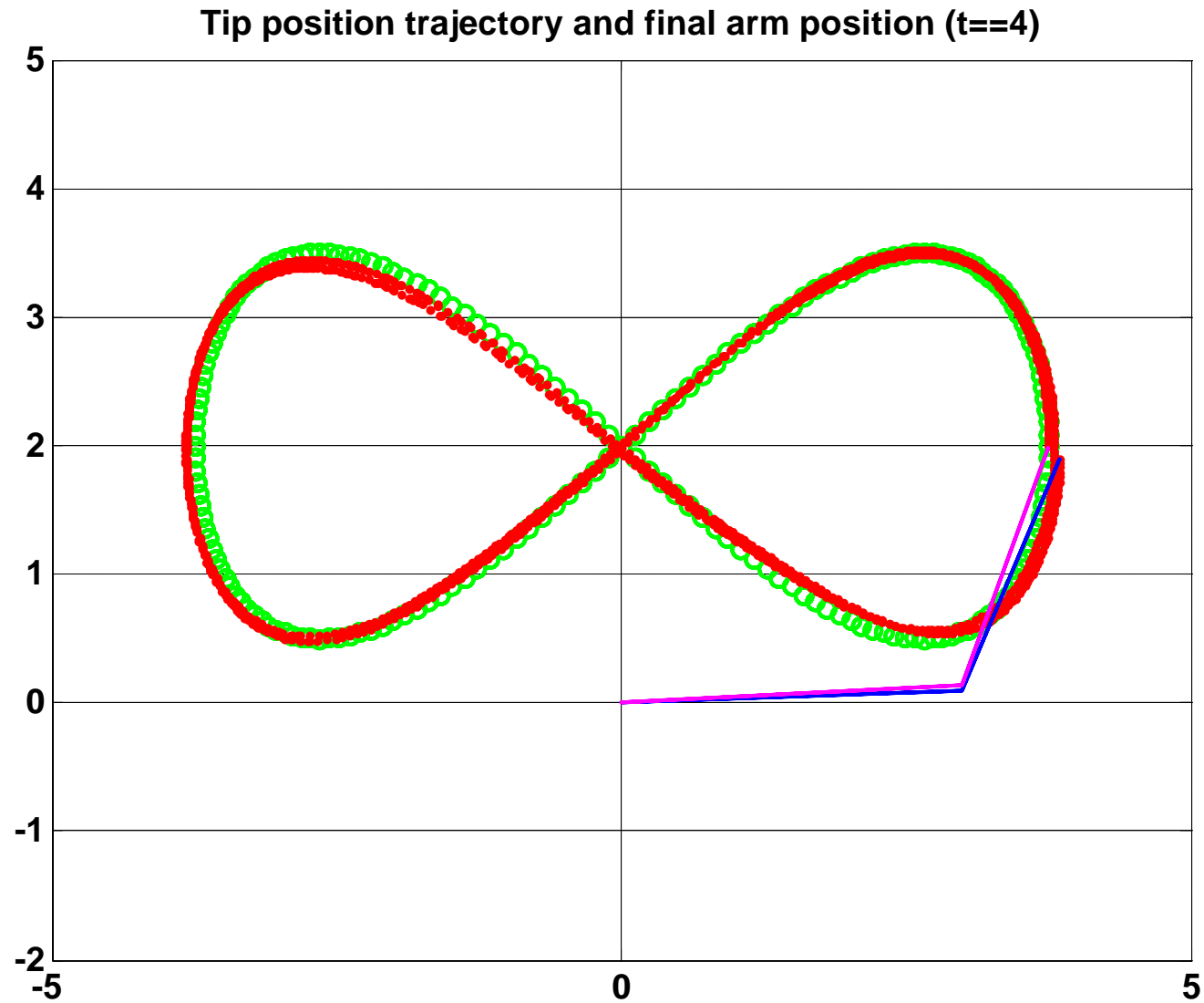


- What's in the Controller box?

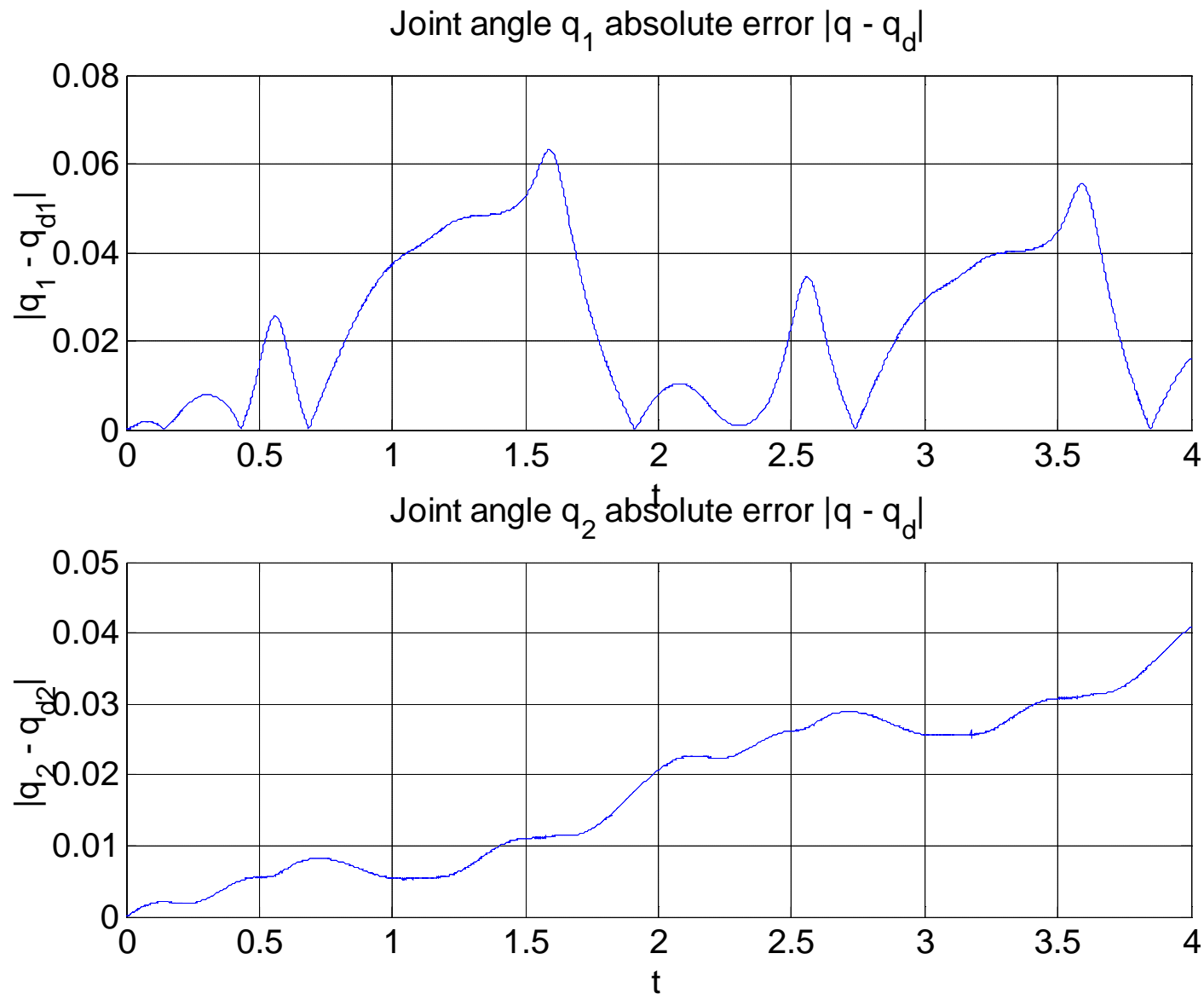
$$u = -K_p(q - q_d) - K_d(\dot{q} - \dot{q}_d) + \dot{q}_d$$

- Problem: The nonlinearities of the system have not been modeled in the controller. We therefore have no guarantee of *asymptotic* tracking of desired trajectories.
- Answer: Use Nonlinear control methods (a bit nasty)
 - Computed Torque Control
 - Passivity Based Control
 - Adaptive Passivity Based Control - here, we separate out the parameters (e.g. masses, lengths) from the robot dynamics and “learn” them as we go with an *additional* controller.

Demo – Linear Joint Control:

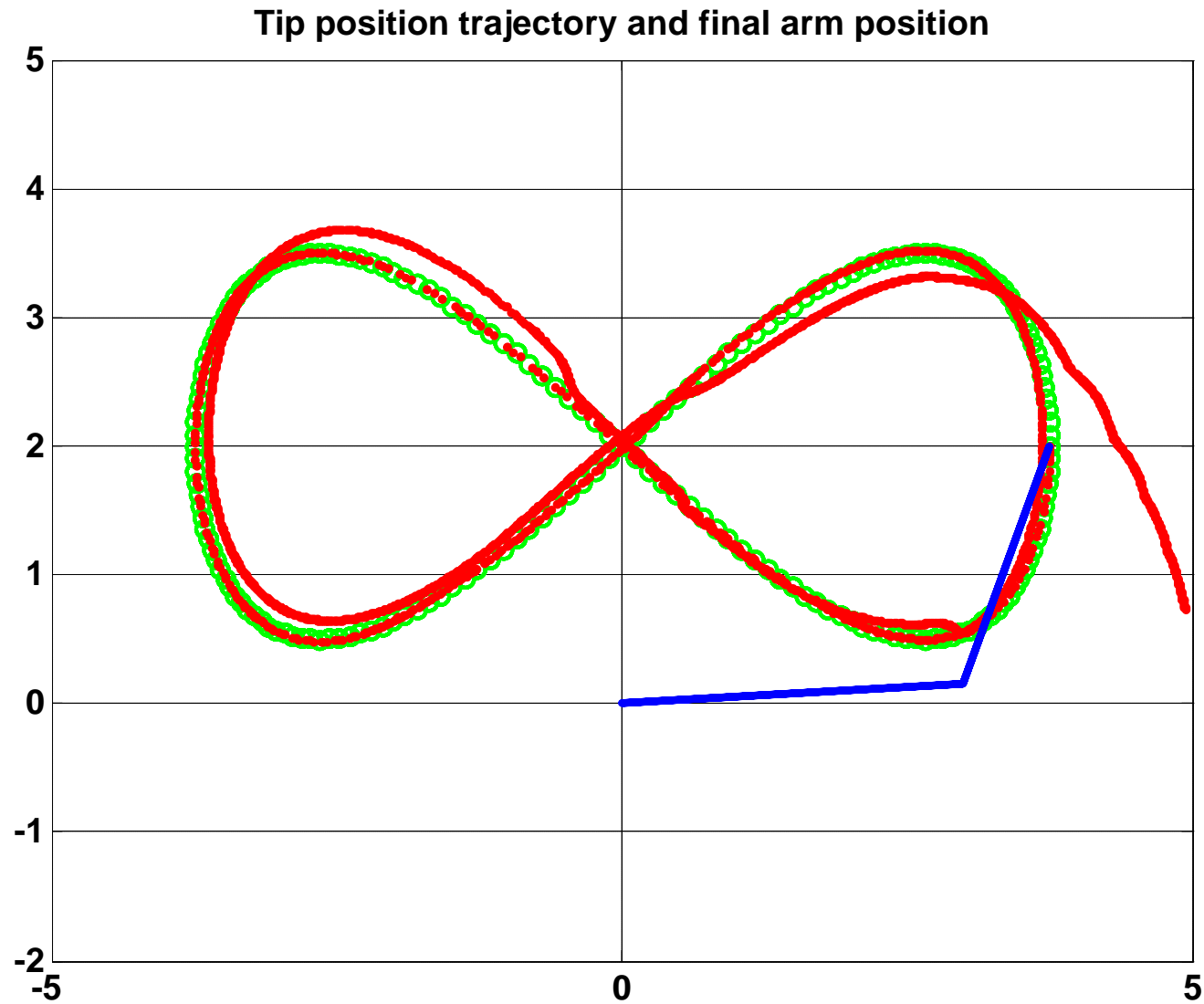


Seems good, but a closer look reveals non-decreasing error:

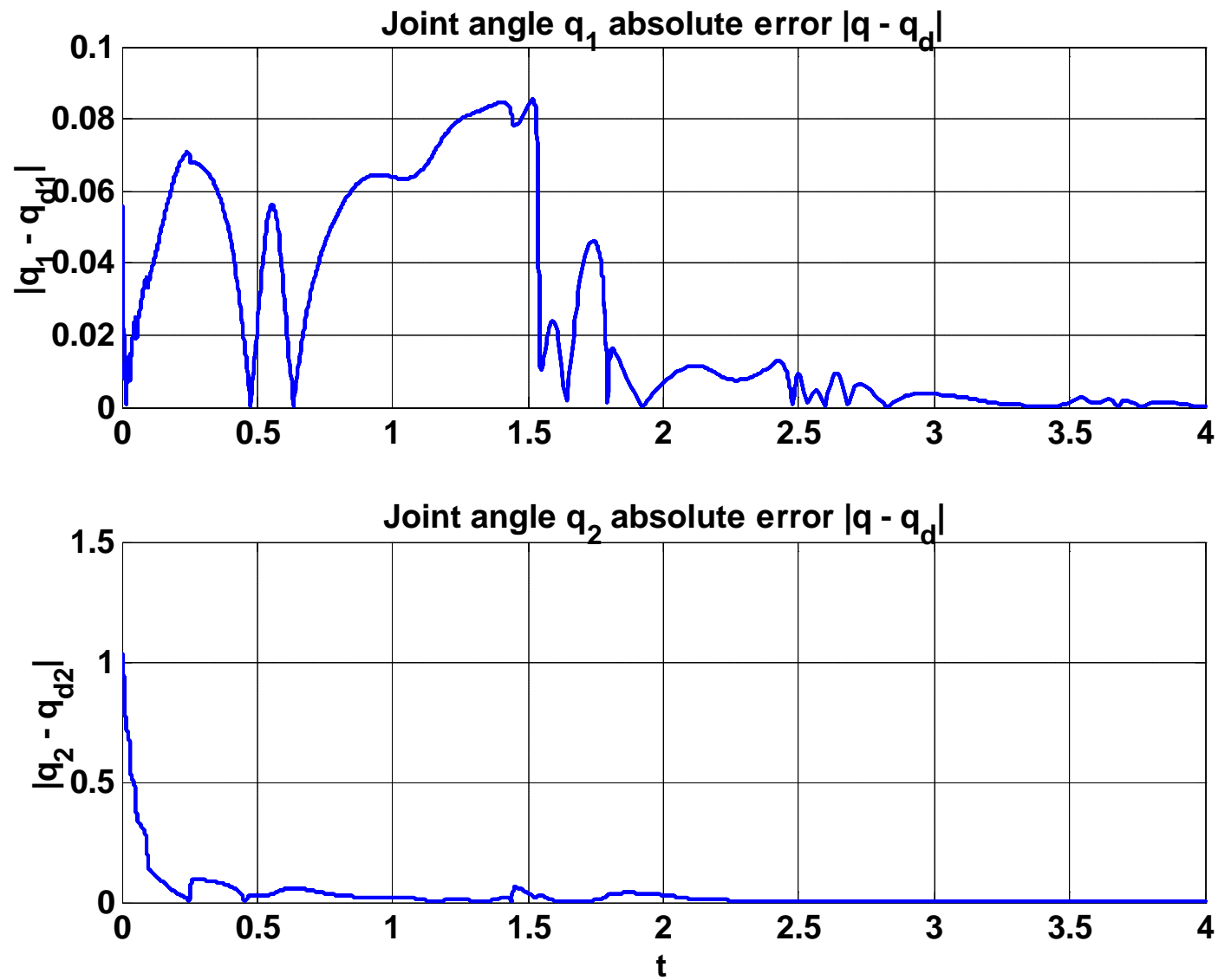


Note that the arm was even started in exactly the right state!

Demo – Adaptive Nonlinear Joint Control:



This is the kind of performance we want:



“Learning” via Adaptive Control

We see the correct mass for each link is “learned” over time:
(they were initialized at the incorrect values of 2 and 4, but converge to the true masses 1 and 1.)

