

ELG7113/MCG5470
MACHINE LEARNING FOR ADAPTIVE AND INTELLIGENT CONTROL SYSTEMS

FINAL EXAM

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Due date posted in Virtual Campus

Policy

The solutions of the assigned problems must be submitted to Virtual Campus as a single file by using the dedicated Assignment. The file must include your name and the assignment number, and ideally it should be a pdf, for example `LastName.FirstName.final.pdf`.

It is acceptable to upload handwritten solutions, but please be sure that your writing is readable. Whenever required, code and outputs have to be attached.

System description

Figure 1 schematizes a mechanical load (pendulum) driven by a DC motor through a gear system, for which non-ideal effects such as backlash can be neglected. The mass M of the pendulum is lumped at the tip, at a distance L from the axis of rotation z which is perpendicular to the plain of motion $\{x, y\}$, and it is parallel to z' , the axis of rotation of the motor's rotor. The absolute rotation θ_L is measured with respect to x , and it related to the motor's shaft rotation θ by the gear relation . The input is the armature voltage E , inducing the armature current I .

The nonlinear governing equations of the system are

$$\dot{\theta} = \omega \quad (1a)$$

$$J\dot{\omega} + D\omega + MgL \frac{N_1}{N_2} \sin\left(\frac{N_1}{N_2}\theta\right) = K_t I \quad (1b)$$

$$L_a \dot{I} + R_a I + K_b \omega = E \quad (1c)$$

where L_a and R_a are the armature's inductance and resistance, K_b is the proportionality parameter between the back electromotive voltage and the angular velocity ω , K_t is the proportionality parameter between motor torque and motor angular velocity, and N_1/N_2 is the gear ratio. The pendulum bar is assumed to be of negligible mass with respect to M . The mechanical inertia and damping are given by $J = J_a + \left(\frac{N_1}{N_2}\right)^2 J_L$ and $D = D_a + \left(\frac{N_1}{N_2}\right)^2 D_L$ in terms of the armature parameters and of the load parameters $J_L = ML^2$ and D_L .

Let (\bar{E}, θ_d) be an equilibrium point corresponding at the constant angle θ_d at the constant armature voltage \bar{E} . By setting the equilibrium conditions in (1), the equilibrium point is

$$\theta = \theta_d, \quad \bar{I} = \frac{\bar{E}}{R_a}, \quad \bar{E} = \frac{R_a N_1}{K_t N_2} MgL \sin\left(\frac{N_1}{N_2}\theta_d\right) \quad (2)$$

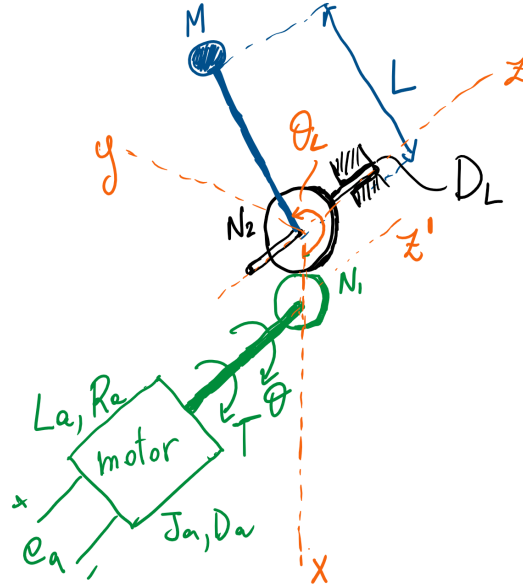


Fig. 1: Electromechanical system.

Define the errors $e_\theta = \theta - \theta_d$, $e_I = I - \frac{\bar{E}}{R_a}$, and $e_\omega = \omega$ (since θ_d is constant). Substitution into (1) give

$$\dot{e}_\theta = \omega \quad (3a)$$

$$J\dot{\omega} + D\omega + MgL\frac{N_1}{N_2} \left(\sin\left(\frac{N_1}{N_2}(e_\theta + \theta_d)\right) - \sin\left(\frac{N_1}{N_2}\theta_d\right) \right) = K_t e_I \quad (3b)$$

$$L_a \dot{e}_I + R_a e_I + K_b \omega = u \quad (3c)$$

where $u = E - \bar{E}$.

Assume the true values of the system's parameters to be $M = 5 \text{ kg}$, $L = 1 \text{ m}$, $D_L = 1 \text{ N m s rad}^{-1}$, $N_1 = 15$ and $N_2 = 45$, $J_a = 2 \text{ kg m}^2 \text{ rad}^{-1}$ and $D_a = 0.1 \text{ N m s rad}^{-1}$. The armature resistance is $R_a = 10 \Omega$ and the inductance is $L_a = 0.1 \text{ H}$. Moreover, the coupling parameters are $K_t = 11$ and $K_b = 6.36$. The coupling parameters are obtained from the motor characteristic torque/angular velocity curve $T + 7\omega = 110$. Let $\theta_d = \frac{\pi}{6}$.

Problem 1

1. **(2pt)** For a constant sampling time step $\Delta = 0.01 \text{ s}$, use the Euler's forward difference scheme to write the error dynamics (3) in discrete time form

$$x_{k+1} = f(x_k) + g(x_k)u_k \quad (4)$$

where k is an integer labelling the discrete time sample. Explicitly identify drift and input dynamics $f(\cdot)$ and $g(\cdot)$.

2. **(20pt)** Consider a quadratic performance index with stage reward $r = x_k^T Q x_k + R u_k^2$, where $Q > 0$ and $R > 0$ are state and control energy weights. Consider a temporal difference learning formulation of the optimal control problem and implement a **value iteration** algorithm to obtain

a control sequence $u_1, u_2 \dots$ with critic function approximation

$$V(x) = \sum_{i=1}^{N_L} \phi_i(x) w_i = \boldsymbol{\phi}^T(x) \mathbf{W} \quad (5)$$

Consider a quadratic homogeneous basis functions set, which determines the dimension of the hidden layer/number of neurons. Simulate the system with initial condition $\theta_0 = 0$, $\omega_0 = 0$ and $I_0 = 0$ (be sure to translate it in terms of errors). Plot the critic gains, the input $u(t) + \bar{E}$ and the trajectory $\theta(t)$.

3. **(20pt)** Repeat the problem at point 2 with the actor function approximation

$$u(x) = \sum_{i=1}^{N_L} \phi_i(x) U_i = \boldsymbol{\phi}^T(x) \mathbf{U} \quad (6)$$

Use the same basis function as the critic approximation. Plot the actor and critic gains, the input u and the trajectory θ .

4. **(20pt)** Let $Q(x_k, u_k) = r(x_k, u_k) + \gamma V(x_{k+1}, u_{k+1})$ be the Q function (not to be confused with the state energy weight in the stage reward). Consider a Q function approximation with homogeneous quadratic basis $\boldsymbol{\psi}(x, u)$ and weights \mathbf{W}

$$Q(x, u) = \sum_{i=1}^{N_L} \psi_i(x, u) w_i = \boldsymbol{\psi}^T(x, u) \mathbf{W} \quad (7)$$

Simulate the system with the same initial conditions and parameters. Plot the Q function gains, the input $u(t)$ and the trajectory $\theta(t)$.

Problem 2

1. **(15pt bonus)** Linearize the discrete time error dynamics around the origin to obtain the linear representation

$$x_{k+1} = Ax_k + Bu_k \quad (8)$$

Clearly identify state and input matrices A and B . Use Q-learning with value function iterations to simulate the system with initial condition $x_0 = (10^\circ, 0, 0)$, and compare the trajectory $\theta(t)$ with the one obtained by solving offline the Riccati equation that gives the sequence of Kalman gains K_k . Pose the offline optimal control as a LQR problem with kernel at the final time S_N to be the identity matrix.