

**Systems Thinking**  
**Mini Project 1**  
**Dynamic Analysis of 2-Link Manipulator**  
**Group: TOPGUN**

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## Question :

- Consider the following system dynamics of a 2-link manipulator :

$$\mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau},$$

$$\mathbf{M} = \begin{bmatrix} M_{11} & M_{12} \\ M_{12} & M_{22} \end{bmatrix}, \mathbf{q} = \begin{bmatrix} q_1 \\ q_2 \end{bmatrix},$$

$$M_{11} = (m_1 + m_2)l_1^2 + m_2l_2(l_2 + 2l_1 \cos(q_2)),$$

$$M_{12} = m_2l_2(l_2 + l_1 \cos(q_2)), M_{22} = m_2l_2^2,$$

$$\mathbf{C} = \begin{bmatrix} -m_2l_1l_2 \sin(q_2) \dot{q}_2 & -m_2l_1l_2 \sin(q_2)(\dot{q}_1 + \dot{q}_2) \\ 0 & m_2l_1l_2 \sin(q_2) \dot{q}_2 \end{bmatrix},$$

$$\mathbf{G} = \begin{bmatrix} m_1l_1g \cos(q_1) + m_2g(l_2 \cos(q_1 + q_2) + l_1 \cos(q_1)) \\ m_2gl_2 \cos(q_1 + q_2) \end{bmatrix}$$

- where  $(m_1, l_1, q_1)$  and  $(m_2, l_2, q_2)$  denote the mass, length and joint angle positions of link 1 and 2 respectively.
- The following parametric values are selected:  $m_1 = 10 \text{ kg}$ ,  $m_2 = 5 \text{ kg}$ ,  $l_1 = 0.2 \text{ m}$ ,  $l_2 = 0.1 \text{ m}$ ,  $g = 9.81 \text{ m/s}^2$ . The joint angles are initially at positions  $[q_1(0)q_2(0)] = [0.1 \ 0.1]$  rad.
- The objective is to bring the the joint angles from the initial position to  $[q_1 \ q_2] = [0 \ 0]$ .
- Q). Via MATLAB simulations (choose P, I, and D gains of your choice) show differences in responses (i.e., plot  $q_1$  vs. t and  $q_2$  vs. t) when (i) PD (ii) PI and (iii) PID controllers are applied separately

**Solution :**

## 2-Link Manipulator: System Dynamics

The system dynamic equation for a 2-link manipulator can be expressed as:

$$M(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = \tau$$

Where:

- $M(q)$  is a  $2 \times 2$  matrix.
- $q$  is a  $2 \times 1$  matrix vector:  $\begin{bmatrix} q_1 \\ q_2 \end{bmatrix}$ .
- $C(q, \dot{q})$  is a  $2 \times 1$  matrix.
- $\dot{q}$  is a  $2 \times 1$  matrix vector:  $\begin{bmatrix} \dot{q}_1 \\ \dot{q}_2 \end{bmatrix}$ .
- $G$  is a  $2 \times 1$  matrix.

All the matrices are given.

The equation can also be represented as:

$$M(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q) = \tau$$

Now, solving for  $\ddot{q}$ , we have:

$$\begin{aligned} &\Rightarrow M(q)\ddot{q} = \tau - [C(q, \dot{q})\dot{q} + G(q)] \\ &\Rightarrow \ddot{q} = M^{-1}(q)[\tau - (C(q, \dot{q})\dot{q} + G(q))] \\ &\Rightarrow \ddot{q} = M^{-1}(q)\tau - M^{-1}(q)[C(q, \dot{q})\dot{q} + G(q)] \longrightarrow [I] \end{aligned}$$

Assuming:

$$\boxed{\hat{\tau} = M^{-1}(q)\tau} \quad (1)$$

Let :

$$\tau = \begin{bmatrix} f_1 \\ f_2 \end{bmatrix}$$

and from equation (1)

$$\tau = \begin{bmatrix} \tau_1 \\ \tau_2 \end{bmatrix} = M(q)\hat{\tau}$$

### **Denoting Error Signals :**

The error signals are denoted as follows:

$$\begin{aligned} e(q_1) &= q_{1f} - q_1 \\ e(q_2) &= q_{2f} - q_2 \end{aligned}$$

Where the target positions of Manipulator Arm 1 and 2 are given by the angles  $q_{1f}$  and  $q_{2f}$ .

### **Initial Positions :**

The initial positions of the system are given as:

$$q_0 = \begin{bmatrix} q_1(0) \\ q_2(0) \end{bmatrix} = \begin{bmatrix} 0.1 \\ 0.1 \end{bmatrix} \rightarrow \text{as provided in the question}$$

### **Modeling PID Group Output :**

The PID group output is modeled as:

$$\begin{aligned} f &= k_p e + k_D \dot{e} + k_I \int e dt \\ f &= \begin{bmatrix} f_1 \\ f_2 \end{bmatrix} \end{aligned}$$

For  $f_1$ :

$$\begin{aligned} f_1 &= k_{p1} e_1(q_1) + k_{D1} \dot{e}_1(q_1) + k_{I1} \int e_1(q_1) dt \\ &= k_{p1}(q_{1f} - q_1) + k_{D1}(\dot{q}_{1f} - \dot{q}_1) + k_{I1} \int (q_{1f} - q_1) dt \\ &= k_{p1}(q_{1f} - q_1) - k_{p1}\dot{q}_1 + k_{I1} \int (q_{1f} - q_1) dt \end{aligned}$$

For  $f_2$ :

$$\begin{aligned} f_2 &= k_{p2} e_2(q_2) + k_{D2} \dot{e}_2(q_2) + k_{I2} \int e_2(q_2) dt \\ &= k_{p2}(q_{2f} - q_2) + k_{D2}(\dot{q}_{2f} - \dot{q}_2) + k_{I2} \int (q_{2f} - q_2) dt \\ &= k_{p2}(q_{2f} - q_2) - k_{p2}\dot{q}_2 + k_{I2} \int (q_{2f} - q_2) dt \end{aligned}$$

### **Actual Torques for the Plant System :**

The actual torques for the plant system are given by:

$$\begin{bmatrix} \tau_1 \\ \tau_2 \end{bmatrix} = M(\theta) \begin{bmatrix} f_1 \\ f_2 \end{bmatrix}$$

### Defining State Variables :

The state variables are defined as follows:

$$\begin{aligned}x_1 &= q_1 \\x_2 &= q_2 \\x_3 &= \dot{q}_1 \\x_4 &= \dot{q}_2 \\x_5 &= \int e_1(q_1) dt \\x_6 &= \int e_2(q_2) dt\end{aligned}$$

### Finding Derivatives of State Variables :

The derivatives of the state variables are given by:

$$\begin{aligned}\dot{x}_1 &= \dot{q}_1 = x_3 \\ \dot{x}_2 &= \dot{q}_2 = x_4 \\ \dot{x}_3 &= \ddot{q}_1 = \phi(x_1, x_2, x_3, x_4, t, x_5, x_6) \\ \dot{x}_4 &= \ddot{q}_2 = \psi(x_1, x_2, x_3, x_4, x_5, x_6, t) \\ \dot{x}_5 &= e_1(q_1) = q_1 f - q_1 = q_1 f - x_1 = 0 - x_1 \\ \dot{x}_6 &= e_2(q_2) = q_2 f - q_2 = q_2 f - x_2 = 0 - x_2\end{aligned}$$

### Final Desired Angles for 2-link manipulator in our dynamics:

The final desired angles for the robot arms are given as:

$$q_{final} = \begin{bmatrix} q_1 f \\ q_2 f \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

From equation [I]:

$$\begin{aligned}\ddot{q} &= M^{-1}(q)\tau - M^{-1}(q)[c(q, \dot{q})\dot{q} + G(q)] \\ &= \hat{\tau} - M^{-1}(q)[c(q, \dot{q})\dot{q} + G(q)] \\ \ddot{q} &= \begin{bmatrix} \ddot{q}_1 \\ \ddot{q}_2 \end{bmatrix}\end{aligned}$$

### State Variables and Initial Conditions :

The state variables are represented as:

$$\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \\ \dot{x}_4 \\ \dot{x}_5 \\ \dot{x}_6 \end{bmatrix} \quad \text{and} \quad x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix}$$

This first order nonlinear differential equation of the form:

$$\dot{x} = \frac{dx}{dt}, \quad x(0) = \text{ initial conditions for our state variables.}$$

Regarding  $x(0)$ , considering the initial conditions as:

$$x(0) = \begin{bmatrix} 0.1 \\ 0.1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Initially, the derivative and integral values of  $q_1$  and  $q_2$  are unknown, and since the system is causal, they are taken initially as zero.

### ODE45 in MATLAB :

In the code, the `ode45` command is used to obtain the state variables  $\bar{x}$  as output from the differential equation vector  $\dot{x}$  of the state variables and the vector representing the initial state as the arguments.

$$[t, s] = \text{ode45}(@(t, \text{state})\text{func}(t, \text{state}), t_{\text{span}}, y_0)$$

Where:

- $y_0$  represents the initial state.
- $\text{func}(t, \text{state})$  : Returns the differentials  $\dot{x}$  vector.

### Main Outputs for Plotting :

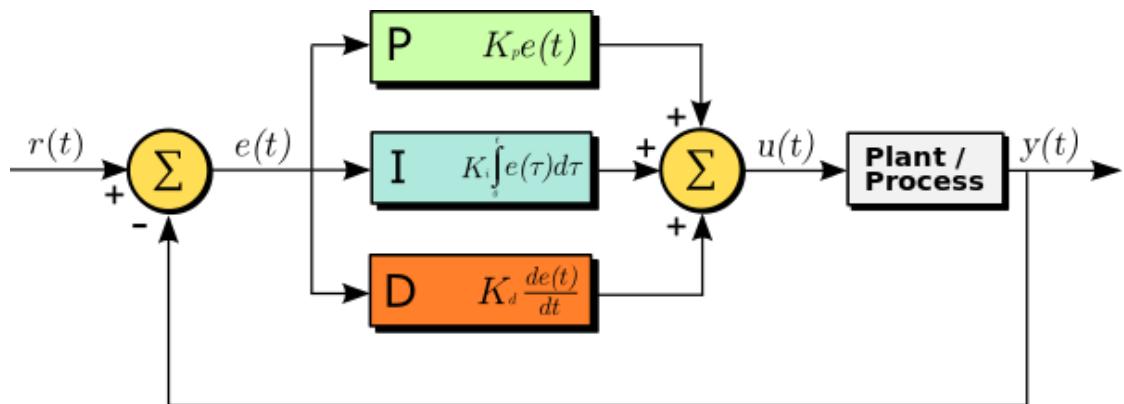
The main outputs to be plotted are  $q_1$  and  $q_2$ , which can be obtained using `ode45`.

$$\begin{aligned} x(1) &= x_1 = q_1 \\ x(2) &= x_2 = q_2 \end{aligned}$$

These values represent the desired angles for the robot arms.

## PID Controller:

A proportional–integral–derivative controller (PID controller or three-term controller) is a control loop mechanism employing feedback that is widely used in industrial control systems and a variety of other applications requiring continuously modulated control. A PID controller continuously calculates an error value  $e(t)$  as the difference between a desired setpoint (SP) and a measured process variable (PV) and applies a correction based on proportional, integral, and derivative terms (denoted P, I, and D respectively), hence the name.



A block diagram of a PID controller in a feedback loop.  $r(t)$  is the desired process value or setpoint (SP), and  $y(t)$  is the measured process value (PV).

## Mathematical form :

The overall control function is given by:

$$u(t) = K_p e(t) + K_i \int_0^t e(\tau) d\tau + K_d \frac{de(t)}{dt}$$

where  $K_p$ ,  $K_i$ , and  $K_d$  (sometimes denoted as  $P$ ,  $I$ , and  $D$ ) are non-negative coefficients for the proportional, integral, and derivative terms, respectively.

## Proportional Term :

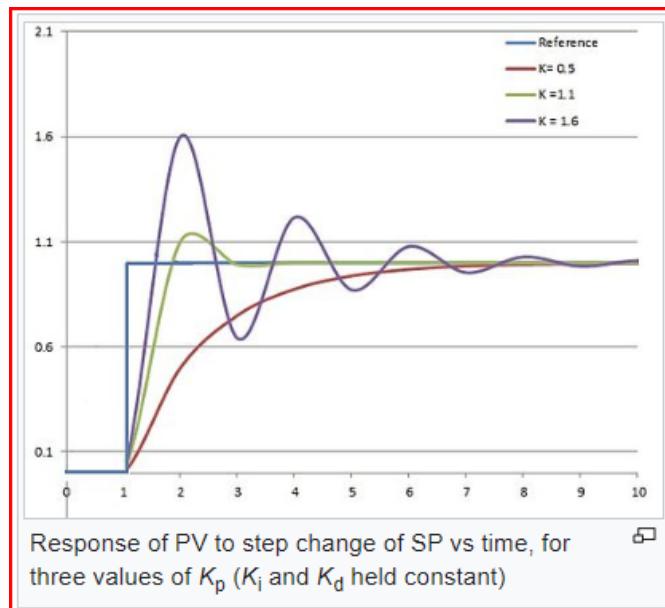
The proportional term produces an output value that is proportional to the current error value. The proportional response can be adjusted by multiplying the error by a constant  $K_p$ , called the proportional gain constant.

The proportional term is given by:

$$P_{\text{out}} = K_p \cdot e(t)$$

In this equation:

- $P_{\text{out}}$  represents the proportional output.
- $K_p$  is the proportional gain constant.
- $e(t)$  represents the current error value.



A high proportional gain results in a large change in the output for a given change in the error. If the proportional gain is too high, the system can become unstable. In contrast, a small gain results in a small output response to a large input error, and a less responsive or less sensitive controller. If the proportional gain is too low, the control action may be too small when responding to system disturbances.

### Integral Term :

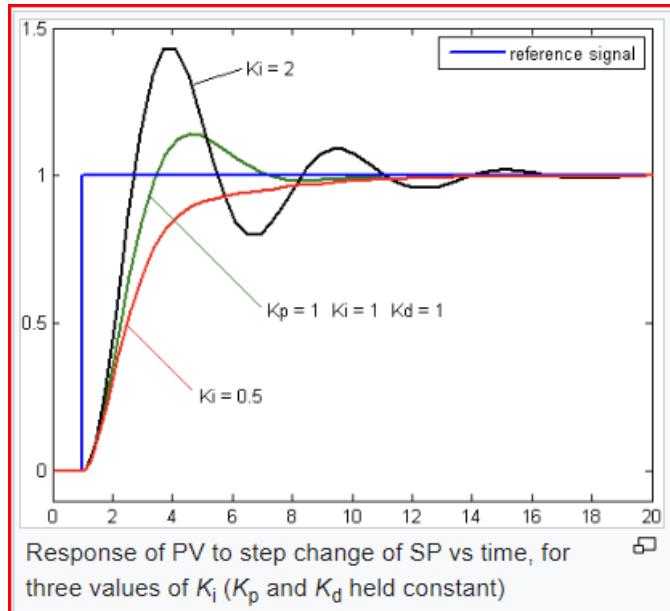
The contribution from the integral term is proportional to both the magnitude of the error and the duration of the error. In a PID controller, the integral term is the sum of the instantaneous error over time and gives the accumulated offset that should have been corrected previously. The accumulated error is then multiplied by the integral gain  $K_i$  and added to the controller output. The integral term is given by:

$$I_{\text{out}} = K_i \int_0^t e(\tau) d\tau$$

Where:

- $I_{\text{out}}$  is the integral term output.
- $K_i$  is the integral gain.

This integral term plays a crucial role in PID control systems, helping to eliminate steady-state errors by accumulating and correcting past errors over time.



The integral term accelerates the movement of the process towards setpoint and eliminates the residual steady-state error that occurs with a pure proportional controller. However, since the integral term responds to accumulated errors from the past, it can cause the present value to overshoot the setpoint value.

## Derivative Term

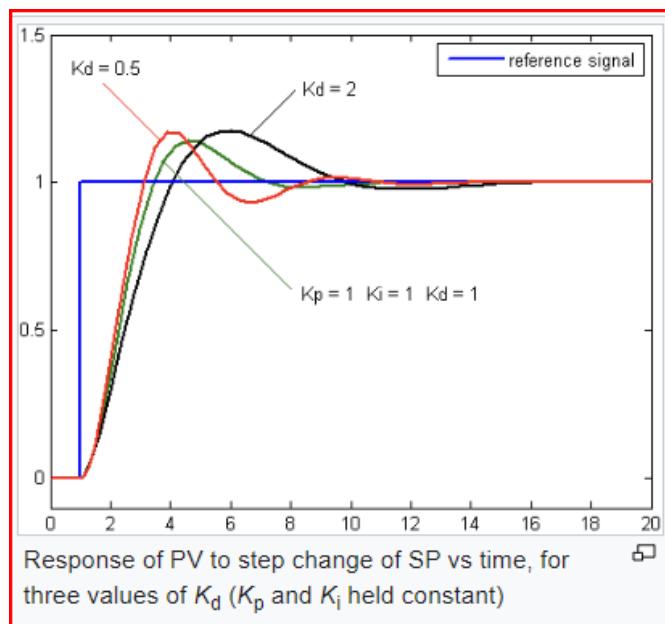
The derivative of the process error is calculated by determining the slope of the error over time and multiplying this rate of change by the derivative gain  $K_d$ . The magnitude of the contribution of the derivative term to the overall control action is termed the derivative gain,  $K_d$ .

The derivative term is given by:

$$D_{\text{out}} = K_d \frac{de(t)}{dt}$$

Where:

- $D_{\text{out}}$  is the derivative term output.
- $K_d$  is the derivative gain.



Derivative action predicts system behavior and thus improves settling time and stability of the system. An ideal derivative is not causal, so that implementations of PID controllers include an additional low-pass filtering for the derivative term to limit the high-frequency gain and noise.

## MATLAB Code :

```

1 clc,clear vars,close all;
2 % Simulation settings
3 t_span = 0:0.01:20;
4 y0 = [0.1, 0.1, 0, 0, 0, 0]; % Initial joint angles and
5 % velocities
6 % Desired joint angles (stop at [theta1_des, theta2_des])
7 theta1_des = 0; % Desired angle for joint 1
8 theta2_des = 0; % Desired angle for joint 2
9
10 % Run simulation
11 [t, s] = ode45(@(t, state) func(t, state, theta1_des,
12 theta2_des), t_span, y0);
13
14 % Plotting
15 figure;
16 subplot(2, 1, 1);
17 plot(t, s(:, 1), 'LineWidth', 2);
18 title('Joint Angle q1 vs Time');
19 xlabel('Time (s)');
20 ylabel('q1 (rad)');
21
22 subplot(2, 1, 2);
23 plot(t, s(:, 2), 'LineWidth', 2);
24 title('Joint Angle q2 vs Time');
25 xlabel('Time (s)');
26 ylabel('q2 (rad)');
27
28 function der_S = func(t, state, theta1_des, theta2_des)
29 % System parameters
30 m1 = 10;
31 m2 = 5;
32 l1 = 0.2;
33 l2 = 0.1;
34 g = 9.81;
35 % Controller gains
36 kp1 = 200;
37 kd1 = 5;

```

```

38    ki1 = 500;
39    kp2 = 400;
40    kd2 = 150;
41    ki2 = 600;
42
43    % Extracting state variables
44    q1 = state(1);
45    q2 = state(2);
46    q1_dot = state(3);
47    q2_dot = state(4);
48    neg_int_q1 = state(5);
49    neg_int_q2 = state(6);
50
51    % Equations of motion
52    m11 = (m1 + m2) * (l1^2) + m2 * l2 * (l2 + 2 * l1 *
53        cos(q2));
54    m12 = m2 * l2 * (l2 + l1 * cos(q2));
55    m22 = m2 * (l2^2);
56
57    M = [m11, m12; m12, m22];
58
59    c11 = -m2 * l1 * l2 * sin(q2) * q2_dot;
60    c12 = -m2 * l1 * l2 * sin(q2) * (q1_dot + q2_dot);
61    c21 = 0;
62    c22 = m2 * l1 * l2 * sin(q2) * q1_dot;
63
64    C = [c11, c12; c21, c22];
65
66    g1 = m1 * l1 * g * cos(q1) + m2 * g * (l2 * cos(q1 +
67        q2) + l1 * cos(q1));
68    g2 = m2 * g * l2 * cos(q1 + q2);
69
70    G = [g1; g2];
71
72    % PDI control law (with desired angles)
73    q1_error = q1 - theta1_des;
74    q2_error = q2 - theta2_des;
75
76    % PDI control law with integral terms
77    tau1 = -kp1 * q1_error - kd1 * q1_dot + ki1 *

```

```

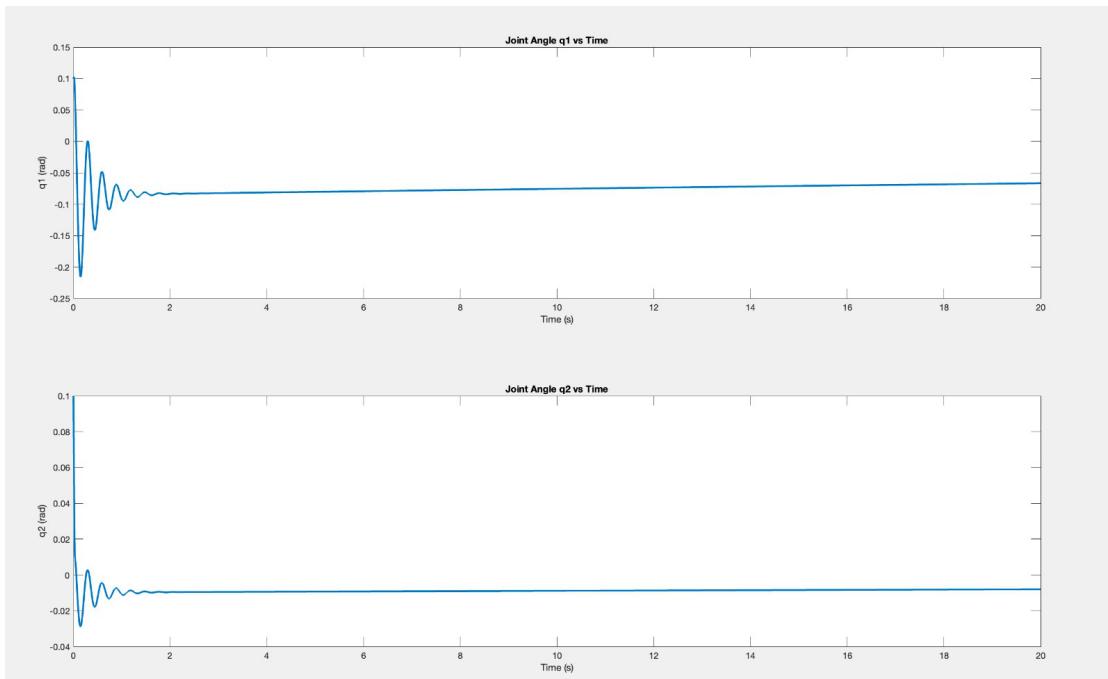
    neg_int_q1;
76 tau2 = -kp2 * q2_error - kd2 * q2_dot + ki2 *
    neg_int_q2;

77 % Control input vector
78 Tau = [tau1; tau2];
80
81 % Solve for q1_ddot and q2_ddot
82 q_ddot = M \ (Tau - C * [q1_dot; q2_dot] - G);
83
84 % State derivatives
85 der_S = [q1_dot; q2_dot; q_ddot(1); q_ddot(2); -
    q1_error; -q2_error];
86 end

```

## Final outputs of joint angles w.r.t time with PID control:

- Case 1 :



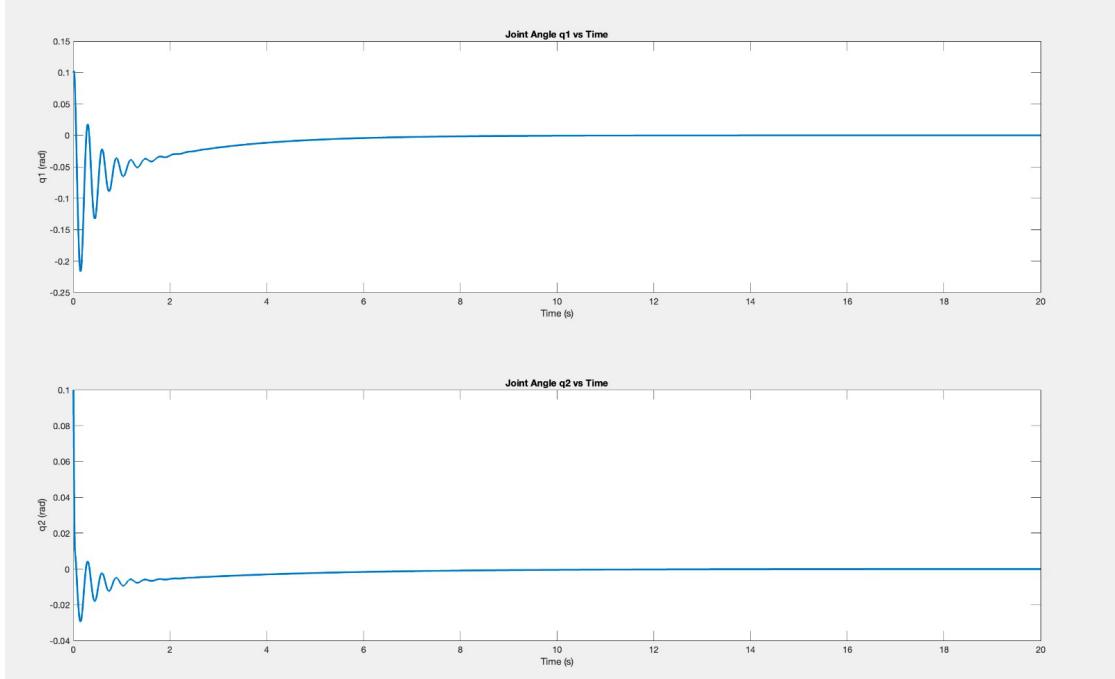
- The values of proportional gains are:  $K_{p1} = 400$ ,  $K_{p2} = 500$ .
- The values of integral gains are:  $K_{i1} = 5$ ,  $K_{i2} = 5$ .
- The values of derivative gains are:  $K_{d1} = 5$ ,  $K_{d2} = 5$ .

### Observations:

1. Oscillatory damping of  $q_1$  and  $q_2$  are observed. The steady-state error values of  $q_1$  and  $q_2$  are non-zero for a timespan of 20 seconds. For example,  $e(q_1) = -0.06$  and  $e(q_2) = -0.008$  at steady state in the given case.
2. We are able to observe noise which is damped with time.

To overcome these above problems in our observation, we need to tune the gains ( $K_p$ ,  $K_d$ , and  $K_i$ ) so as to achieve steady state faster, avoid noise, and reduce steady-state error.

- Case 2 :

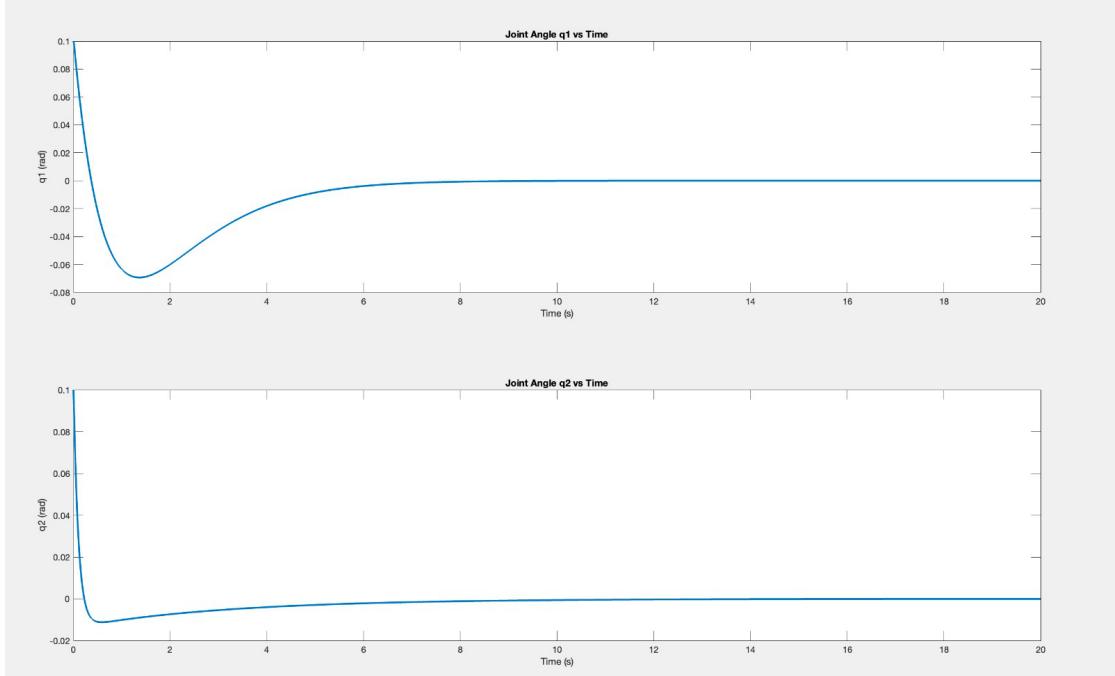


- The integral gains  $K_{i1}$  and  $K_{i2}$  have been increased with the goal of achieving a setpoint of 0 as the system reaches steady state.
- $\Rightarrow e(q_1)$  and  $e(q_2)$  approach approximately 0 at steady state.
- The values of integral gains are:  $K_{i1} = 200$ ,  $K_{i2} = 150$ .
- The values of derivative gains are:  $K_{d1} = 5$ ,  $K_{d2} = 5$ .
- We observe that the proportional and derivative gains are kept constant, while the integral gains have been increased:  
 $K_{i1}$  changed from 5 to 200,  
 $K_{i2}$  changed from 5 to 150.
- Since the steady-state error of  $q_2$  was comparatively less than that of  $q_1$ , the integral gain of  $q_1$  was increased to a larger extent than that of  $q_2$ .

#### **Observation:**

Now we observe that the steady-state errors are close to zero. Our aim now is to reduce the noise before achieving steady state, i.e., we need to achieve steady state quickly within our specified time span.

- Case 3 :



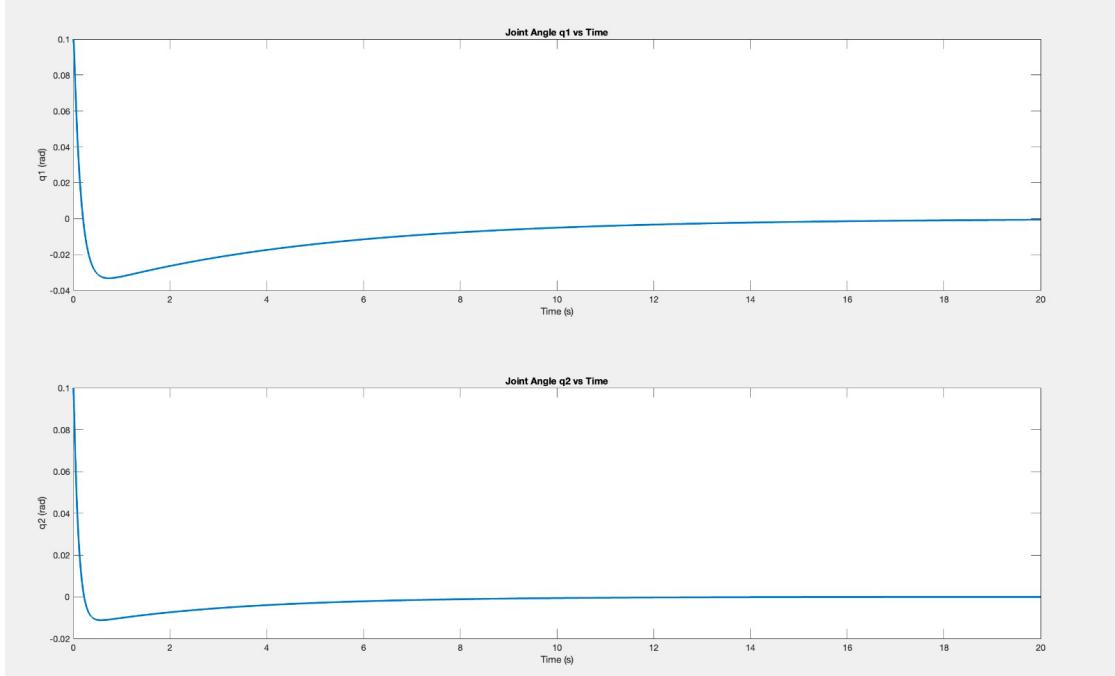
- In order to reduce oscillations, the derivative gains  $K_{d1}$  and  $K_{d2}$  are increased.
- $q_1$  exhibits rapid oscillations of larger amplitude compared to  $q_2$ . Therefore,  $K_{d1}$  is increased to a larger value than  $K_{d2}$ .
- The values of proportional gains are:  $K_{p1} = 400$ ,  $K_{p2} = 500$ .
- The values of integral gains are:  $K_{i1} = 200$ ,  $K_{i2} = 150$ .
- The values of derivative gains are:  $K_{d1} = 200$ ,  $K_{d2} = 50$ .
- $K_{d1}$  has been increased from 5 to 200, and  $K_{d2}$  has been increased from 5 to 50.

#### Observations:

The oscillations have reduced, but steady state is achieved in more time for  $q_1$  (approximately 7 seconds) within our time span taken. Our aims are as follows:

1. To make  $q_1$  achieve steady state in less time (leading to case 4).
2. To avoid the initial variation of  $q_1$  about the setpoint when it reaches the setpoint for the first time.

- Case 4 :

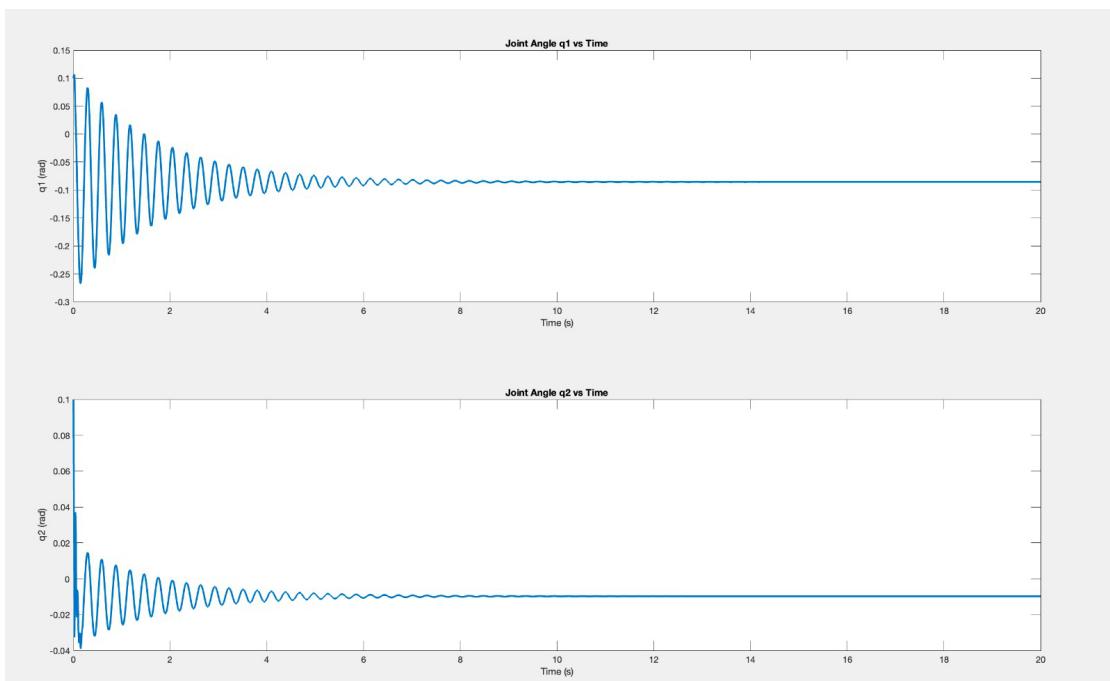


- To further reduce the variation of  $q_1$  about the setpoint and ensure  $q_1$  achieves the setpoint in less time, we have adjusted the controller gains. Specifically, we reduced the derivative gain and increased the proportional gain.
  - The values of proportional gains are:  $K_{p1} = 1000$ ,  $K_{p2} = 500$ .
  - The values of integral gains are:  $K_{i1} = 200$ ,  $K_{i2} = 150$ .
  - The values of derivative gains are:  $K_{d1} = 150$ ,  $K_{d2} = 50$ .
  - Changes:  
 $K_{p1}$  was increased from 400 to 1000  $K_{d1}$  was reduced from 200 to 150.
  - The derivative term generally slows down the rate of change of  $q_1$  with respect to time, which increases the time taken to achieve steady state. Therefore,  $K_{d1}$  was reduced to decrease the time taken to achieve steady state. On the other hand,  $K_{p1}$  was increased to enhance the rate at which  $q_1$  drops from 0.1 rad to setpoints.
  - As a result, the required goal of PID control was achieved, which includes the reduction of oscillations and reaching the given setpoint (0 radians) within the specified time span of 20 seconds.

**Final outputs of joint angles w.r.t time with PD control:**

In PD controller, we take the integral gain  $K_i = 0$

- Case 1 :



- The values of proportional gains are:  $K_{p1} = 400$ ,  $K_{p2} = 500$ .
- The values of integral gains are:  $K_{i1} = 0$ ,  $K_{i2} = 0$ .
- The values of derivative gains are:  $K_{d1} = 1$ ,  $K_{d2} = 1$ .
- Initially, we set our derivative gains as low as possible (here, 1) to demonstrate that  $q_1$  and  $q_2$  exhibit oscillations up to a certain time interval before reaching steady state.

#### Observations:

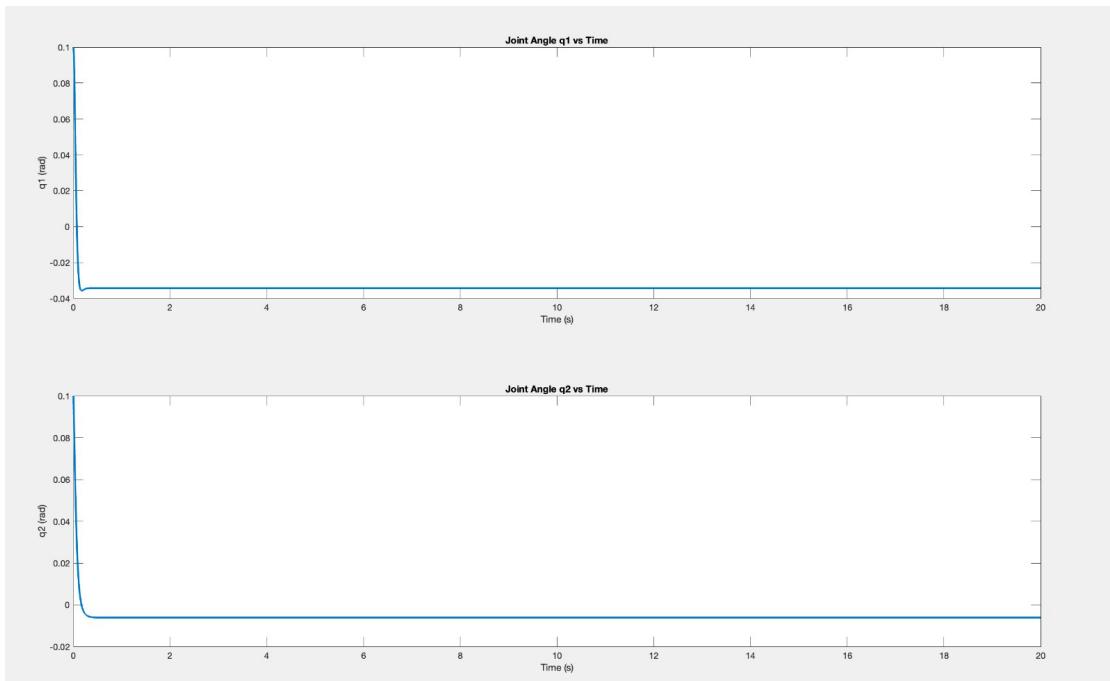
1. Oscillations are observed.

2. The steady-state errors of  $q_1$  and  $q_2$  are non-zero. Typically, the integral term accumulates the error from previous states, multiplied by time, to help the joint angles achieve the setpoint with approximately zero steady-state error.

**Aim:**

To reduce oscillations and decrease the time taken to achieve steady state.

- **Case 2:**



- The values of proportional gains are:  $K_{p1} = 1000$ ,  $K_{p2} = 800$ .
- The values of integral gains are:  $K_{i1} = 0$ ,  $K_{i2} = 0$ .
- The values of derivative gains are:  $K_{d1} = 50$ ,  $K_{d2} = 50$ .
- To fasten the process and reduce the settling time, the following adjustments were made:
  - Proportional gains were increased:  
 $K_{p1}$ : 400 → 1000  
 $K_{p2}$ : 500 → 800

- Derivative gains were increased to reduce oscillations around the setpoint before reaching steady state:  
 $K_{d1}: 1 \rightarrow 50$   
 $K_{d2}: 1 \rightarrow 50$

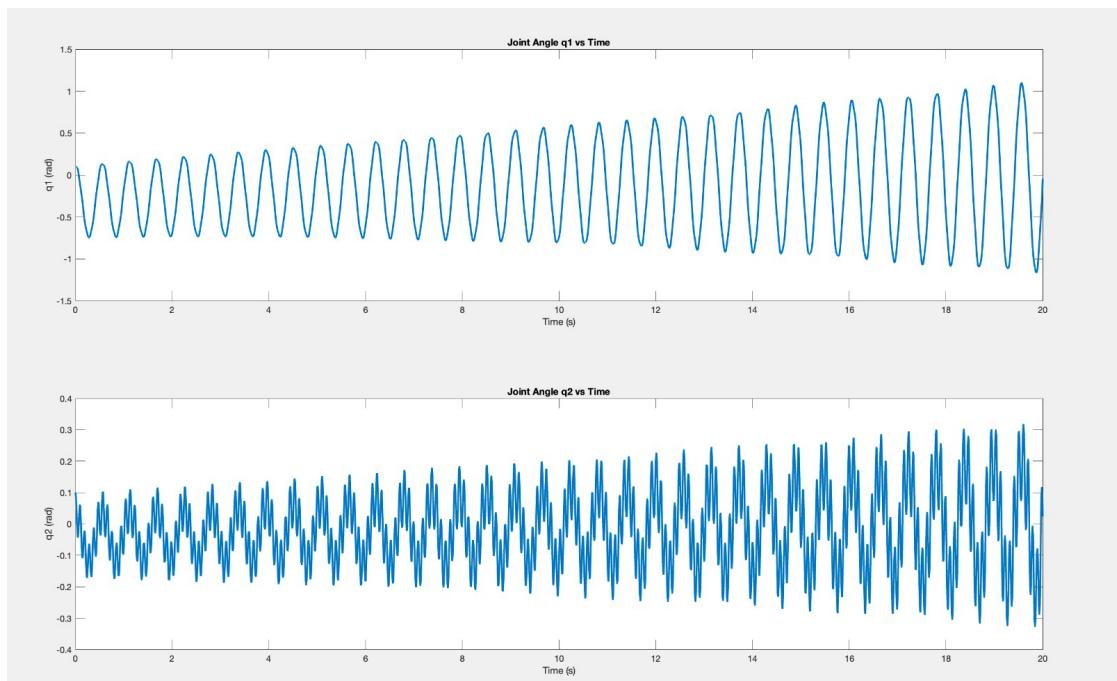
### Observations:

Steady states were achieved with some error (integral gains  $K_{i1}$  and  $K_{i2}$  set to 0). For these gains, the steady-state errors are approximately:

- $e(q_1) = -0.03$
- $e(q_2) = -0.06$

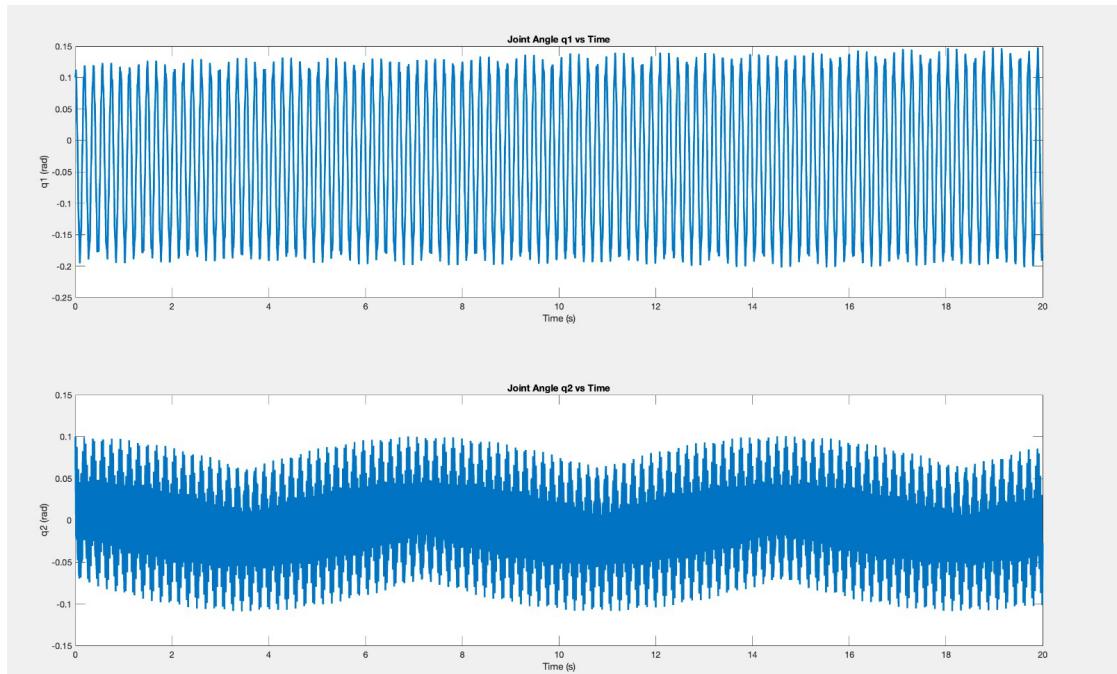
## PI Controller :

- Case 1 :



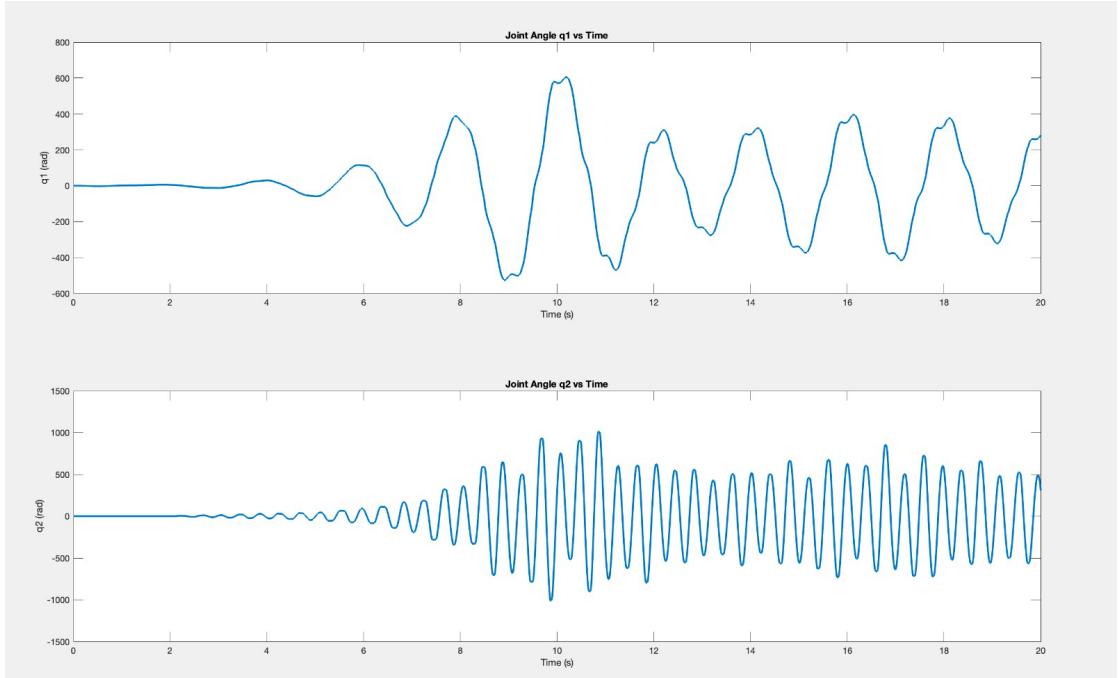
- The values of proportional gains are:  $K_{p1} = 100$ ,  $K_{p2} = 100$ .
- The values of integral gains are:  $K_{i1} = 10$ ,  $K_{i2} = 10$ .
- The values of derivative gains are:  $K_{d1} = 0$ ,  $K_{d2} = 0$ .

- Case 2 :



- The values of proportional gains are:  $K_{p1} = 1000$ ,  $K_{p2} = 1000$ .
- The values of integral gains are:  $K_{i1} = 10$ ,  $K_{i2} = 10$ .
- The values of derivative gains are:  $K_{d1} = 0$ ,  $K_{d2} = 0$ .

- Case 3 :



- The values of proportional gains are:  $K_{p1} = 5$ ,  $K_{p2} = 10$ .
- The values of integral gains are:  $K_{i1} = 10$ ,  $K_{i2} = 10$ .
- The values of derivative gains are:  $K_{d1} = 0$ ,  $K_{d2} = 0$ .

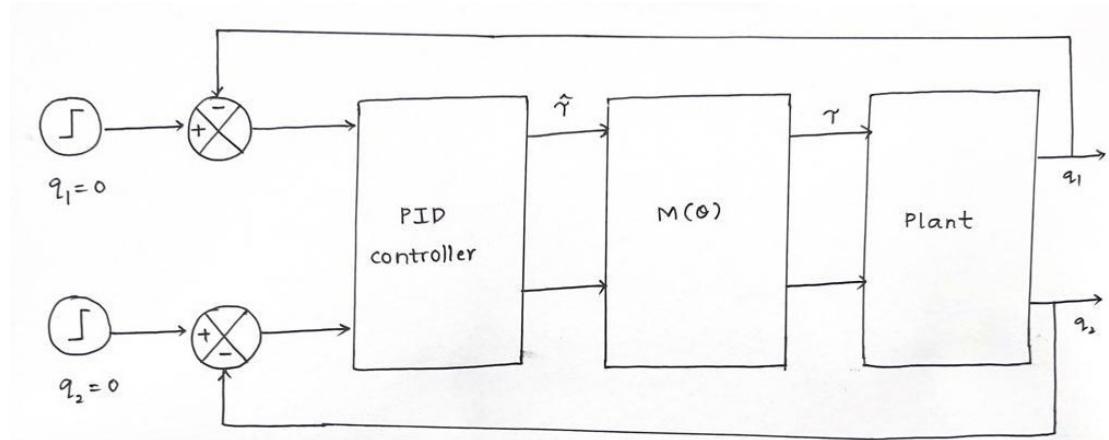
### Observations :

- In all three cases, oscillations are observed for any time span considered, and our joint angles do not reach a steady state.
- In a PI (Proportional-Integral) controller, the integral term accumulates the error over time.
- When the system starts, the integral term accumulates error quickly because there's no damping from the derivative term ( $K_d = 0$ ).
- This can lead to what's called "integral windup," where the integral term becomes very large and causes the system to overshoot and oscillate.

- Without the derivative term, the system can become underdamped, meaning it oscillates around the desired setpoint.
- The proportional and integral terms alone might not provide enough damping to bring the system to a stable state.

## Simulink:

To perform the same in simulink we have obtained a block diagram as shown :



To implement a model in Simulink, we have obtained the differential equations for the chosen state variables.

Previously, we demonstrated that  $\ddot{q}$  can be derived from the equations:

$$\ddot{q} = M^{-1}(q)\tau - M^{-1}(q)[c(q, \dot{q})\dot{q} + G(q)]$$

The exact equations for  $\ddot{q}$  are:

$$\mathbf{M}'(\theta) [\ddot{\theta} + G] + \mathbf{M}'(\theta)\tau =$$

$$\begin{aligned}
& \left[ \begin{aligned}
& -m_2^2 l_1 l_2^3 \sin x_2 (2x_3 \dot{x}_4 + x_4^2) + m_1 m_2 l_1 l_2^2 g \cos x_1 + m_2^2 g l_2^3 \cos(x_1 + x_2) \\
& + m_2 l_2^1 \cos x_1 - m_2^2 l_1 l_2 \sin x_2 x_4^2 (L_2 + l_1 \cos x_2) - m_3 g l_2^2 \cos(x_1 + x_3) (L_2 + l_1) \\
& m_2 l_1 l_2 \sin x_2 (L_2 + l_1 \cos x_2) (2x_3 \dot{x}_4 + x_4^2) - m_2 l_1 l_2 g \cos x_1 (L_2 + l_1 \cos x_2) \\
& - m_2^2 g l_2 \cos(x_1 + x_2) (L_2 + l_1 \cos x_2) - m_2^2 g l_1 l_2 \cos x_1 (L_2 + l_1 x_2) \\
& + (m_1 + m_2) L_2 l_1 l_2 \sin x_2 x_4^2 + (m_1 + m_2) L_2 m_2 g l_2 \cos(x_1 + x_2) \\
& + m_2 l_1 l_2 \sin x_2 x_4^2 (L_2 + l_1 \cos x_2) + m_2 g l_2^2 \cos(x_1 + x_2) (L_2 + l_1 \cos x_1)
\end{aligned} \right]
\end{aligned}$$

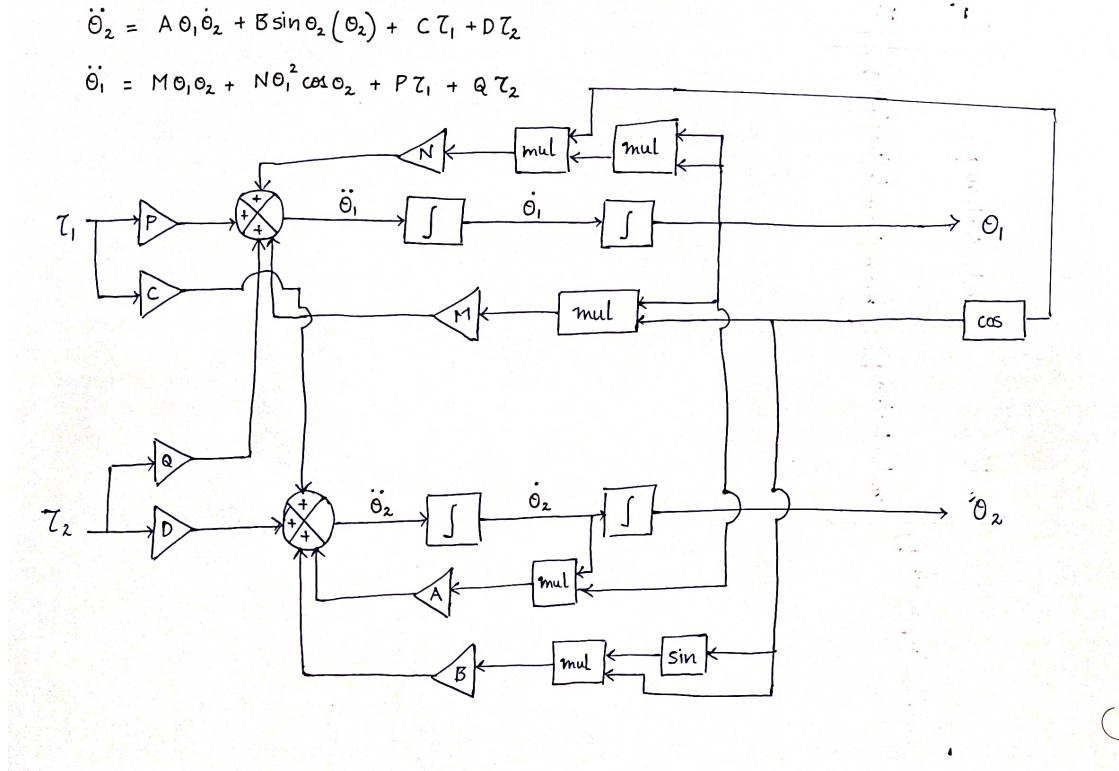
$$= \begin{aligned}
& \left[ \begin{aligned}
& m_1 m_2 l_1 l_2^2 g \cos x_1 + m_2^2 g l_2^2 \cos x_1 \cos x_2 (x_1 + x_2) - m_2^2 g l_2 \cos x_2 \cos(x_1 + x_2) \\
& - 2m_2 l_1 l_2^3 x_3 x_4 \sin x_2 x_2 \\
& - m_2^2 l_1 l_2^2 \sin x_2 x_4^2 \cos x_2 - 2m_2 l_1 l_2^3 \sin x_2 x_4 \\
& + m_2 g l_1 l_2^2 \cos x_2 \cos(x_1 + x_2) + 2m_2 l_1 l_2 \sin x_2 (L_2 + l_1 \cos x_2) x_4^2 \\
& + m_2 l_1 l_2 \sin x_2 x_3 x_4 (L_2 + l_1 \cos x_2) + (m_1 + m_2) l_1 l_2 m_2 g \cos(x_1 + x_2) \\
& + (m_1 + m_2) l_1^3 m_2 l_3 \sin x_2 x_4^2 - m_2 g l_1 l_2 \cos x_1 (L_2 + l_1 \cos x_2) (m_1 + m_2)
\end{aligned} \right]
\end{aligned}$$

$$\mathbf{M}^{-1}(\theta) \cdot \mathbf{T} = \quad (2)$$

$$\frac{1}{m_2 L_1^2 L_2^2 (m_1 - m_2 \sin^2(q))} \begin{bmatrix} m_2 L_2^2 & -m_1 L_2 (L_2 + L_1 \cos q_2) \\ -m_2 L_2 (L_2 + L_1 \cos q_2) & m_1^2 L_1^2 + m_1 L_1^2 + m_2 L_2 (L_2 + 2L_1 \cos q_2) \end{bmatrix} \begin{bmatrix} T_1 \\ T_2 \end{bmatrix} \quad (3)$$

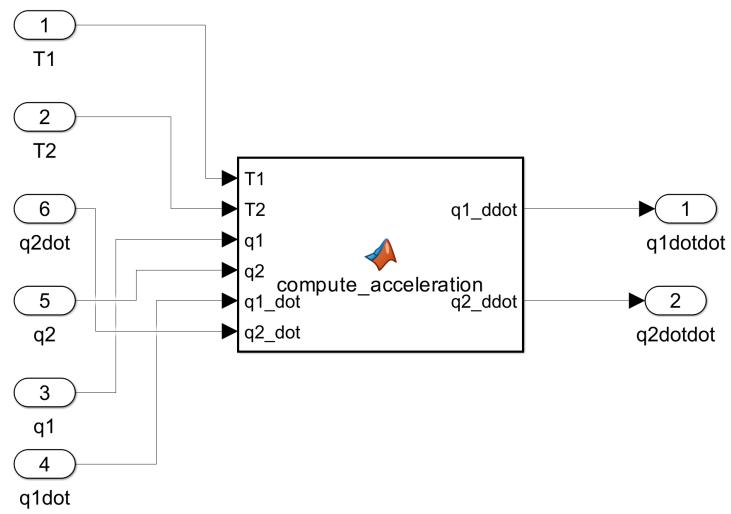
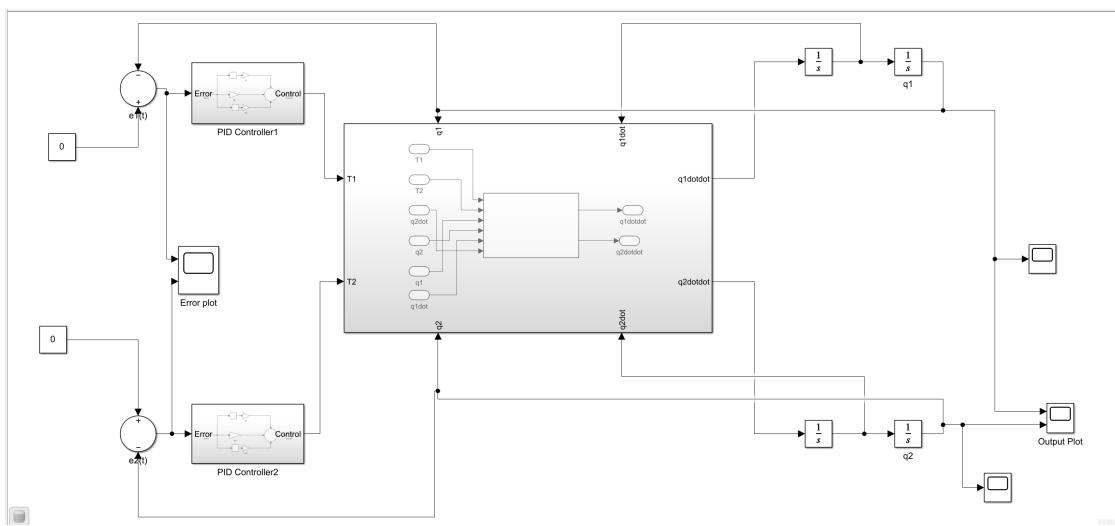
$$\ddot{q} = M^{-1}(q)\tau - M^{-1}(q)[c(q, \dot{q})\dot{q} + G(q)]$$

The output  $\ddot{q}$  from this equation can be used to model the plant in Simulink.

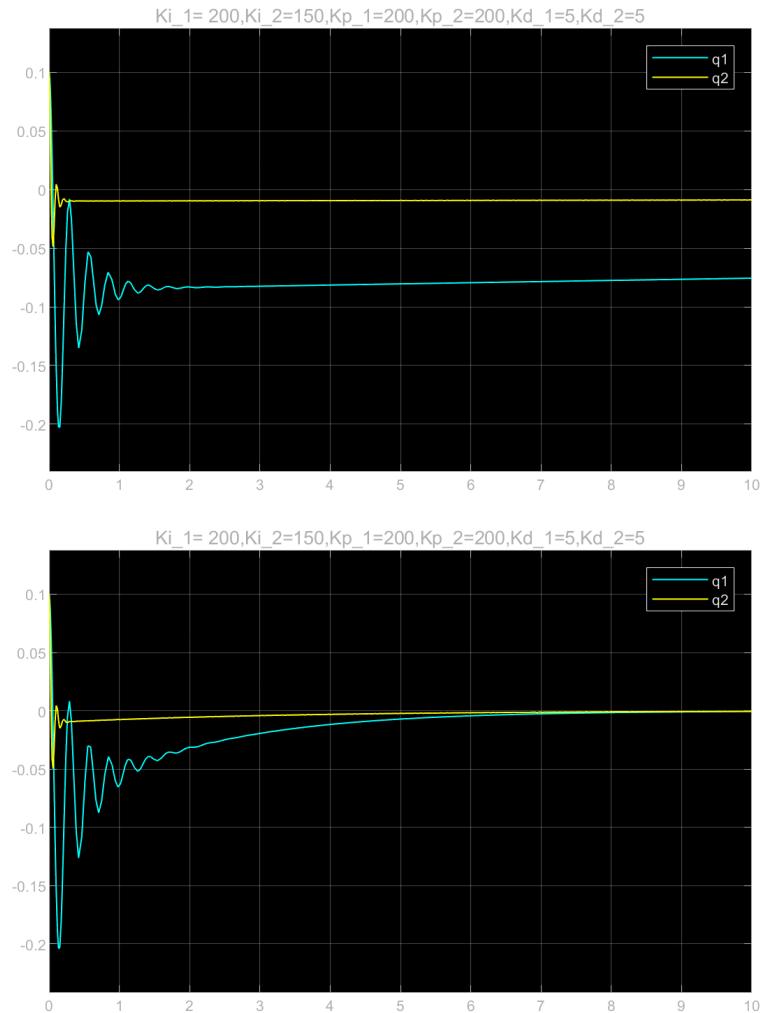


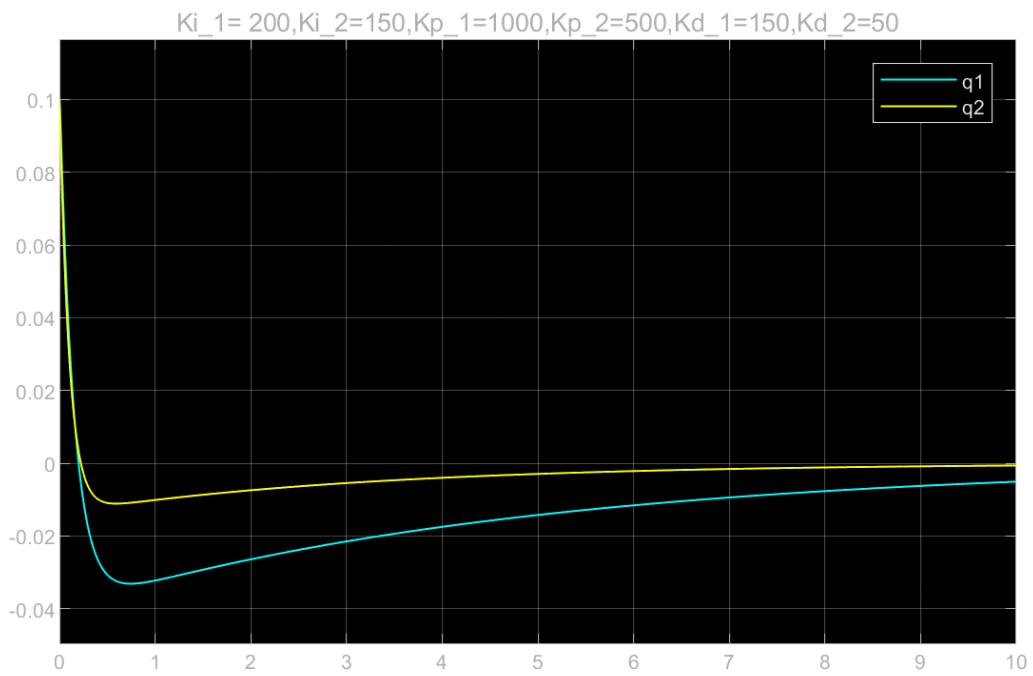
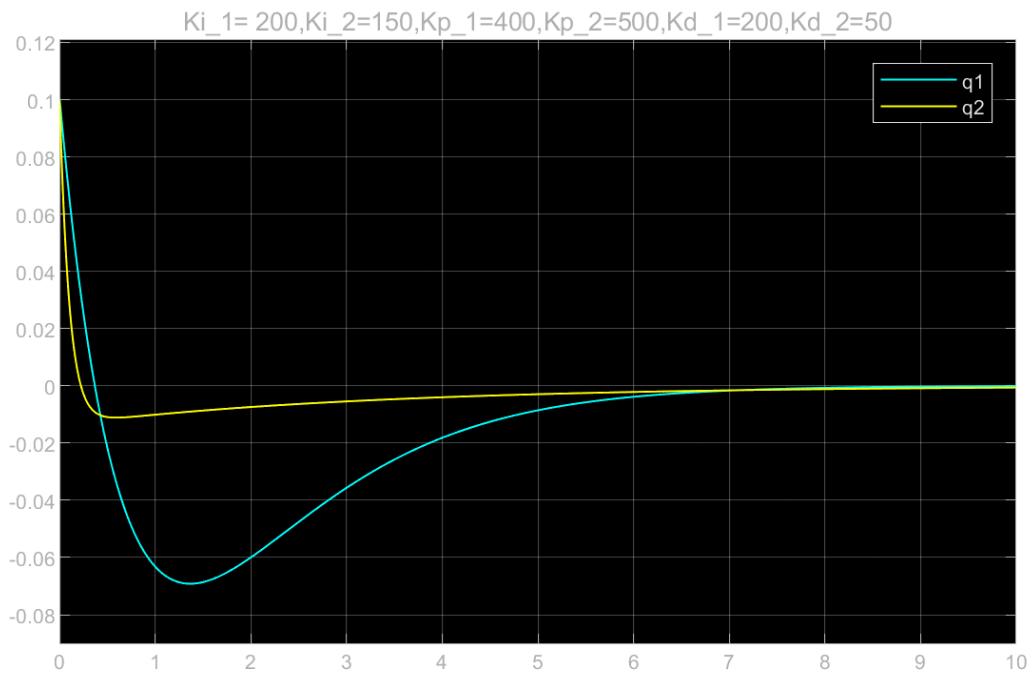
and the plant can be implemented as similar to the above in simulink.

1. Initially, the inputs  $r_1(t)$  and  $r_2(t)$  (joint angles) are set to 0 rad.
2. The final outputs obtained from the plant are the joint angles obtained for that particular instant, and these  $q_1$  and  $q_2$  are fed back as unity feedback to calculate errors at that instant as shown in the figure.
3. The corresponding errors at that instant are sent into the PID controllers where the controllers are tuned according to our needs.
4. The outputs of the controllers are the angular accelerations  $\ddot{q}$ , which are further multiplied by the moment of inertia  $M(q)$  to obtain the torques, which are the actual inputs to be given to the plant.
5. The plant computes our joint angles from the torques and the state variables taken ( $q_d$  and  $\ddot{q}$ ) as shown above.
6. So, finally, our system brings the joint angles to their steady states by estimating the errors in the joint angles at every instant of time and generating the torques from the PID, PD, PI controllers tuned respectively.



# Exploring PID Controller Efficiency: Simulink Results for Diverse Tuning Parameters





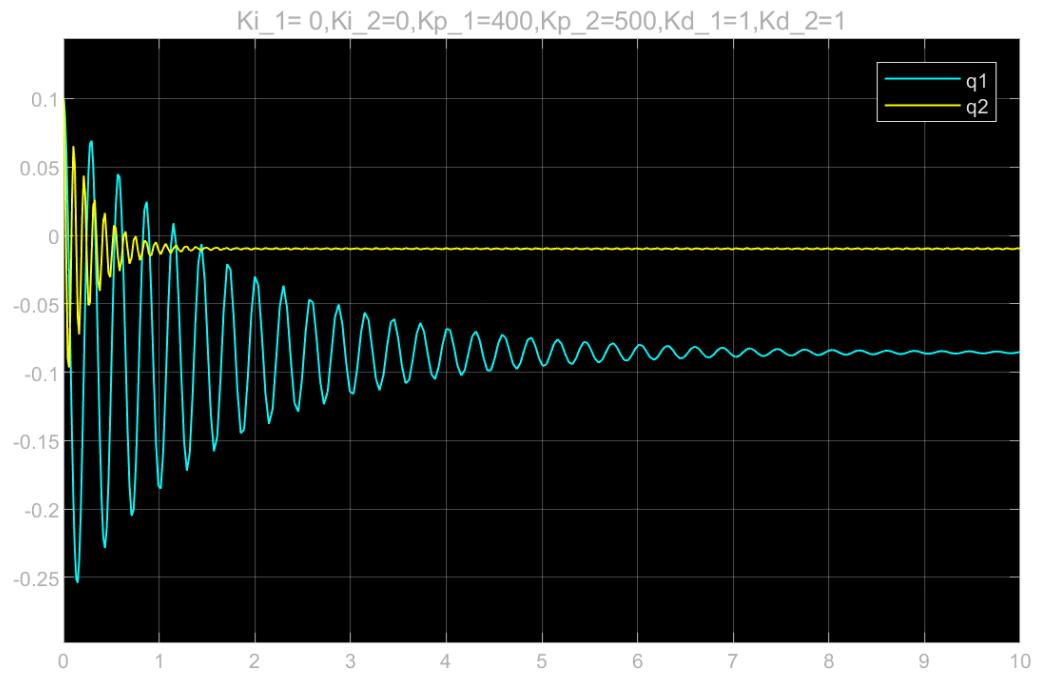
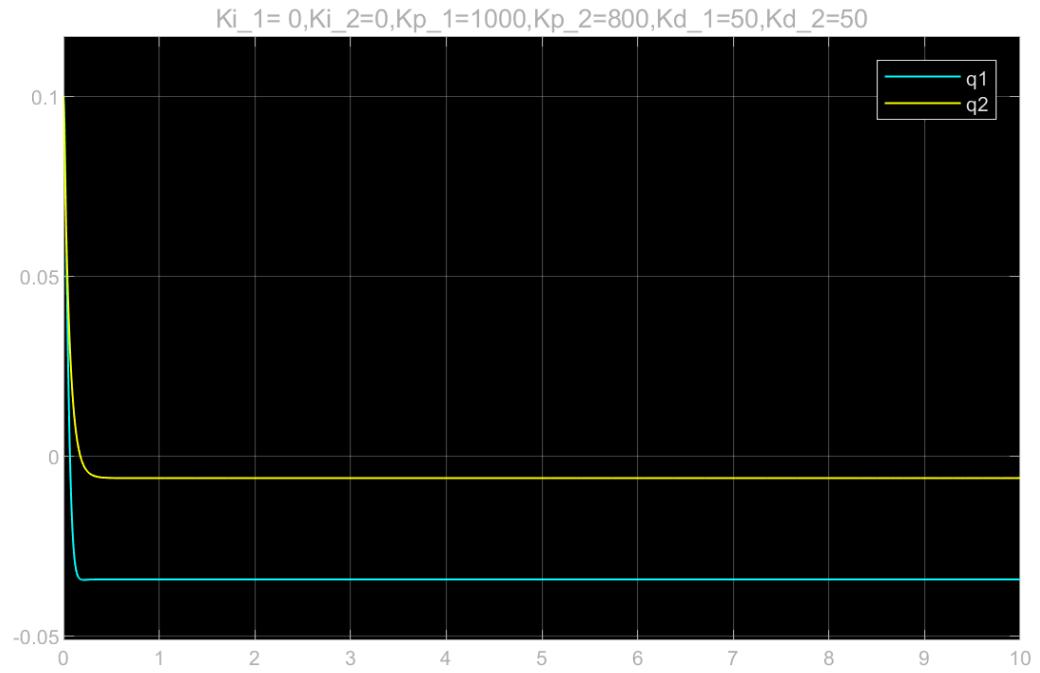
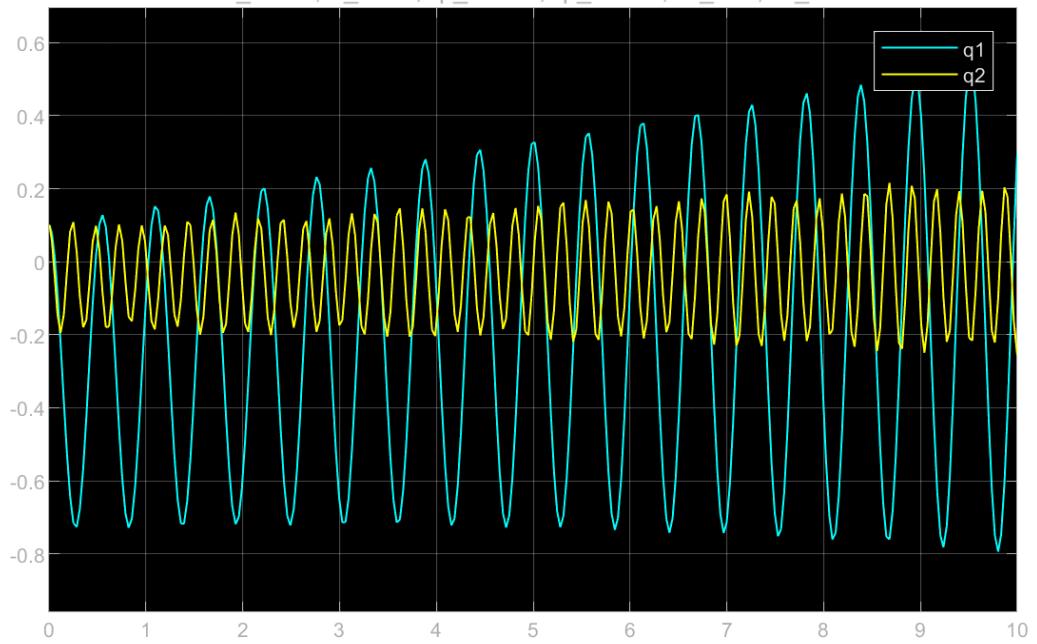


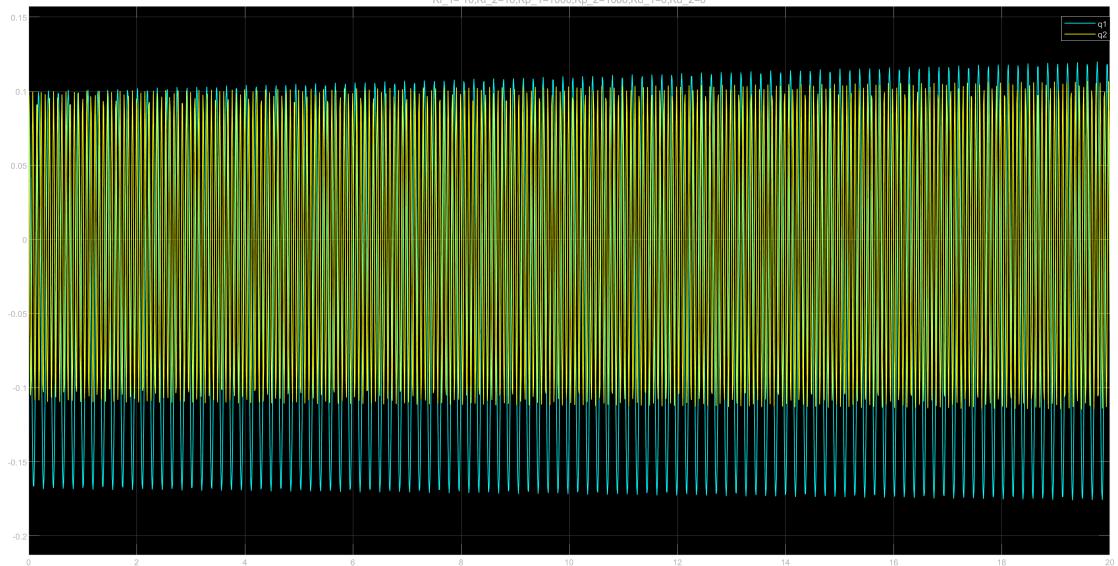
Figure 1: Caption for PD1



$Ki\_1 = 10, Ki\_2 = 10, Kp\_1 = 100, Kp\_2 = 100, Kd\_1 = 0, Kd\_2 = 0$



$Ki\_1 = 10, Ki\_2 = 10, Kp\_1 = 1000, Kp\_2 = 1000, Kd\_1 = 0, Kd\_2 = 0$



# Effect of KD, KP, and KI Values on System Behavior

## A. Effect on Damping

Aspect	High KD	Low KD
Effect on Damping	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Effective at damping oscillations</li> <li>- Smoother and more stable control</li> <li>- Less prone to vibrations, ideal for precise movements</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Can lead to over-damping, slow response</li> <li>- Reduces performance, lacks responsiveness in dynamic movements</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- More responsive system</li> <li>- Faster movement, agile and suited for rapid tasks</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Less effective damping</li> <li>- Prone to overshoot, reducing stability and accuracy</li> </ul>

Table 1: Effect of KD on Damping

## B. Reduction of Overshoot

Aspect	High KD	Low KD
Reduction of Overshoot	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Reduces overshoot, provides smooth control</li> <li>- Essential for precise, controlled movements</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Over-damping may delay reaching the desired position</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Faster initial response, tolerates some overshoot</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Inadequate reduction of overshoot, may affect precision</li> </ul>

Table 2: Reduction of Overshoot with KD

## C. Effect on Response Time

Aspect	High KD	Low KD
Effect on Response Time	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Faster settling time, quick stability</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Excessive KD can cause overshoot or slow response</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Rapid initial response, slightly extended settling time</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Persistent oscillations may slow the overall settling process</li> </ul>

Table 3: Effect of KD on Response Time

## D. Stability

Aspect	High KD	Low KD
<b>Stability</b>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- High stability, reduces overshoot and oscillations</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Over-damping can slow down response, potentially affecting system sensitivity</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- More responsive, reduces risk of over-damping</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Ineffective damping can cause instability and unpredictable behavior</li> </ul>

Table 4: Effect of KD on Stability

## E. Conclusion

In practice, tuning the derivative gain (KD) involves finding the right balance between reducing overshoot, damping oscillations, and maintaining responsiveness based on the specific characteristics and requirements of our robotic arm control system. The optimal KD value will depend on the following factors: mechanical properties of the system, desired performance, and the specific tasks it needs to perform.

## F. Effect of Proportional Gain (KP)

Aspect	High KP	Low KP
<b>Effect of KP on Error Correction</b>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Strong response to errors</li> <li>- Faster convergence to desired angles</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Overshooting, oscillations, and instability</li> <li>- Increased sensitivity to noise</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Stable and less aggressive response</li> <li>- Smoother, more predictable system</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Slow convergence, delays in achieving targets</li> </ul>

Table 5: Effect of KP on System Behavior

## G. Effect of Integral Gain (KI)

Aspect	High KI	Low KI
<b>Effect of KI on Steady-State Error</b>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Eliminates steady-state errors</li> <li>- Enhances accuracy</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Risk of instability and oscillations</li> <li>- Requires careful tuning</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- More stable response</li> <li>- Reduces sensitivity to noise and disturbances</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Persistent steady-state errors</li> </ul>

Table 6: Effect of KI on System Behavior

## H. Steady-State Error Reduction

Aspect	High KI	Low KI
<b>Steady-State Error Reduction</b>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Eliminates steady-state errors</li> <li>- Precise control without long-term deviations</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Risk of instability with excessive KI</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Stable response but allows some steady-state error</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Does not effectively eliminate steady-state errors</li> </ul>

Table 7: Steady-State Error Reduction with KI

## I. Conclusion

In practice, tuning the PI controller involves finding the right combination of KP and KI to achieve the desired control performance while avoiding instability, overshooting values, or oscillations. The optimal values will depend on the specific characteristics of our robotic arm system and the control requirements of your application (refer Simulink).

## J. Effect of Proportional Gain (KP)

Aspect	High KP	Low KP
<b>Pros</b>	<ul style="list-style-type: none"> <li>- Reduces steady-state error quickly by scaling control error based on current error</li> <li>- Ensures high accuracy and precision, resulting in fine control</li> <li>- Allows precise positioning of the robotic arm</li> </ul>	<ul style="list-style-type: none"> <li>- Provides a more stable response to errors</li> <li>- Less likely to cause oscillations and instability</li> </ul>
<b>Cons</b>	<ul style="list-style-type: none"> <li>- Can lead to oscillations and instability, making it harder to achieve precise control</li> <li>- May cause overshooting and erratic behavior due to excessive control action</li> </ul>	<ul style="list-style-type: none"> <li>- May result in slower error reduction due to a less aggressive control system</li> <li>- Increased time to reach the desired position</li> </ul>

Table 8: Effect of Proportional Gain (KP) on System Behavior

## K. Effect of Derivative Gain (KD)

Aspect	High KD	Low KD
<b>Effect of KD on System Response</b>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Dampens oscillations, provides smooth control</li> <li>- High stability when correctly tuned</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Can lead to an over-damped response, slow settling</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Rapid initial response</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Ineffective at damping oscillations</li> </ul>

Table 9: Effect of KD on System Behavior

## L. Effect of Integral Gain (KI)

Aspect	High KI	Low KI
<b>Effect of KI on System Response</b>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Aggressive response to long-term errors</li> <li>- Accurate response to persistent disturbances</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Can lead to instability and unpredictability in noisy environments</li> </ul>	<p><b>Pros:</b></p> <ul style="list-style-type: none"> <li>- Smooth, gradual control response</li> <li>- Reduced risk of oscillations and instability</li> </ul> <p><b>Cons:</b></p> <ul style="list-style-type: none"> <li>- Persistent errors may remain</li> </ul>

Table 10: Effect of KI on System Behavior

## M. Conclusion

In practice, tuning the PID controller involves finding the right combination of KP , KI , and KD to achieve the desired control performance while avoiding instability, over- shooting values, or oscillations. The optimal values will depend on the specific characteristics of our robotic arm system and the control requirements of your application (refer Simulink).