

## Assignment 4

By- Aishwary Jagetia

### Contents

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- Notations: For a given variable,  $x$ ,  $\dot{x}$  is its time derivative,  $\ddot{x}$  is
- create symbolic variable for  $x$ .
- specify your initial and final condition.
- Implement the Augmented PD control for set point tracking.
- Augmented PD Control

### Notations: For a given variable, $x$ , $\dot{x}$ is its time derivative, $\ddot{x}$ is

---

2nd-order derivative.

```
clc
clear all;
close all;
% the following parameters for the arm
I1=10; I2 = 10; m1=5; r1=.5; m2=5; r2=.5; l1=1; l2=1; g=9.8;

% we compute the parameters in the dynamic model
a = I1+I2+m1*r1^2+ m2*(l1^2+ r2^2);
b = m2*l1*r2;
d = I2+ m2*r2^2;
```

### create symbolic variable for $x$ .

---

$x_1$  -  $\theta_1$   $x_2$  -  $\theta_2$

```
symx= sym('symx',[4,1]);

global M C Gq
M = [a+2*b*cos(symx(2)), d+b*cos(symx(2));
     d+b*cos(symx(2)), d];
C = [-b*sin(symx(2))*symx(4), -b*sin(symx(2))*(symx(3)+symx(4)); b*sin(symx(2))*symx(3),0];
invM = inv(M);
invMC= inv(M)*C;

% Gravity Matrix
g1=-(m1+m2)*g*l1*sin(symx(2))-m2*g*l2*sin(symx(1)+symx(2));
g2=-m2*g*l2*sin(symx(1)+symx(2));
Gq=[g1;g2];
```

### specify your initial and final condition.

---

```
x0= [-0.5,0.2,0.1,0.1];
global w torque
w=0.2;
tf=10;
torque = [];
```

### Implement the Augmented PD control for set point tracking.

---

```

xf = [0, 0, 0, 0];
options = odeset('RelTol',1e-4,'AbsTol',[1e-4, 1e-4, 1e-4, 1e-4]);
[T,X] = ode45(@(t,x) AugmentedPDControl(t,x),[0 tf],x0, options);

figure('Name','Theta_1 under PD SetPoint Control');
plot(T, X(:,1),'r-');
hold on
plot(T, w*ones(size(T,1),1),'b-');
figure('Name','Theta_2 under PD SetPoint Control');
plot(T, X(:,2),'r--');
hold on
plot(T, sin(2*T),'b-');

figure('Name','Augmented PD Control');
plot(T, torque(1,1:size(T,1)),'-');
hold on
plot(T, torque(2,1:size(T,1)),'r--');
hold off
torque=[];

```

## Augmented PD Control

```

function dx = AugmentedPDControl(t,x)
    w=0.2;
    theta_d=[w;sin(2*t)]; % [x1d;x2d] Desired trajectory
    dtheta_d=[0;2*cos(2*t)]; % [x1d_dot;x2d_dot]
    ddtheta_d=[0;-4*sin(2*t)]; % [x1d_ddot;x2d_ddot]
    theta=x(1:2,1); % [x1;x2]=[x(1);x(2)]
    dtheta=x(3:4,1); % [x1_dot;x2_dot]=[x(3);x(4)]

    global M C Gq M_d C_d Gq_d
    symx= sym('symx',[4,1]);
    M = subs(M, [symx(1);symx(2);symx(3);symx(4)], [x(1);x(2);x(3);x(4)]);
    C = subs(C, [symx(1);symx(2);symx(3);symx(4)], [x(1);x(2);x(3);x(4)]);
    Gq = subs(Gq, [symx(1);symx(2);symx(3);symx(4)], [x(1);x(2);x(3);x(4)]);
    M_d = subs(M, [symx(1);symx(2);symx(3);symx(4)], [theta_d(1);theta_d(2);dtheta_d(1);dtheta_d(2)]);
    C_d = subs(C, [symx(1);symx(2);symx(3);symx(4)], [theta_d(1);theta_d(2);dtheta_d(1);dtheta_d(2)]);
    Gq_d = subs(Gq, [symx(1);symx(2);symx(3);symx(4)], [theta_d(1);theta_d(2);dtheta_d(1);dtheta_d(2)]);
    invM = inv(M);
    invMC= inv(M)*C;

    tau = Controler(theta_d,dtheta_d,ddtheta_d,theta,dtheta);

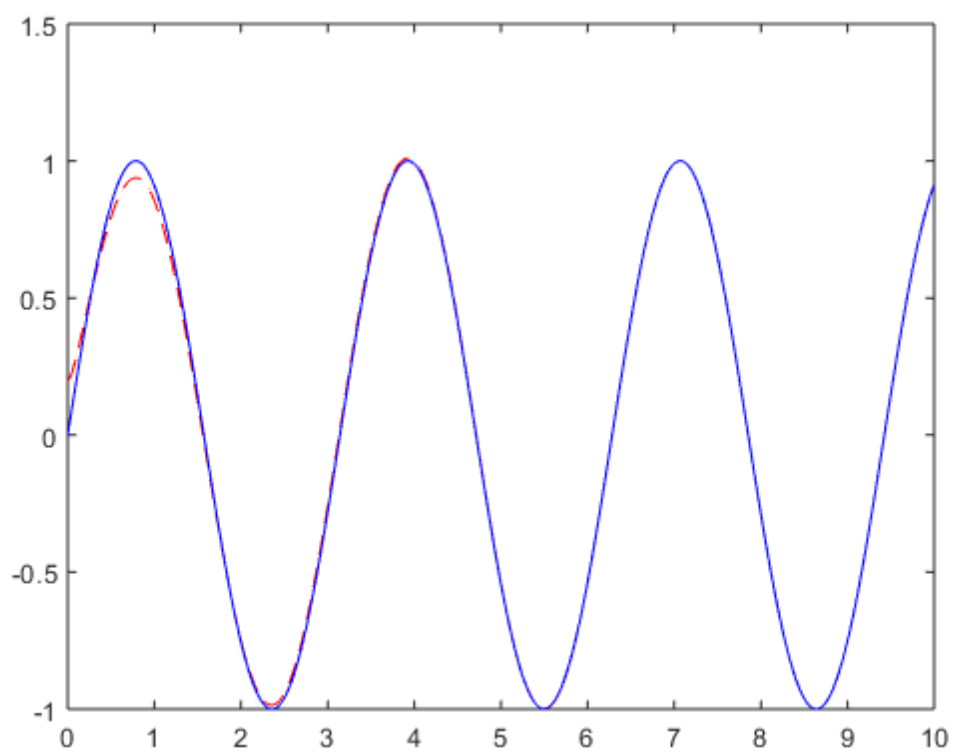
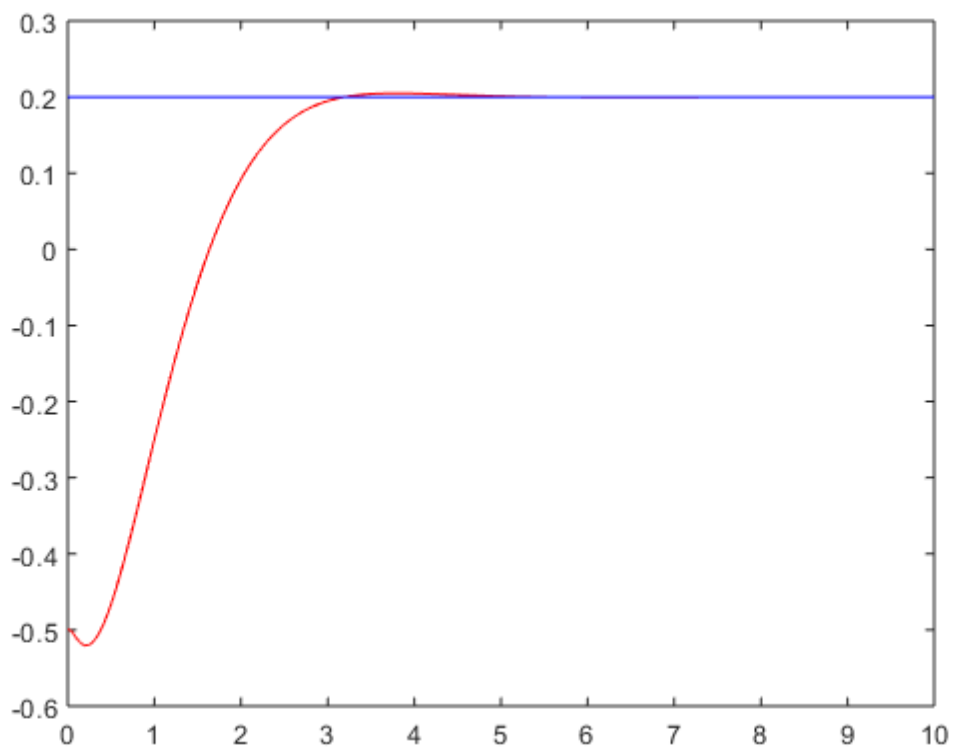
    global torque
    torque = [torque, tau];

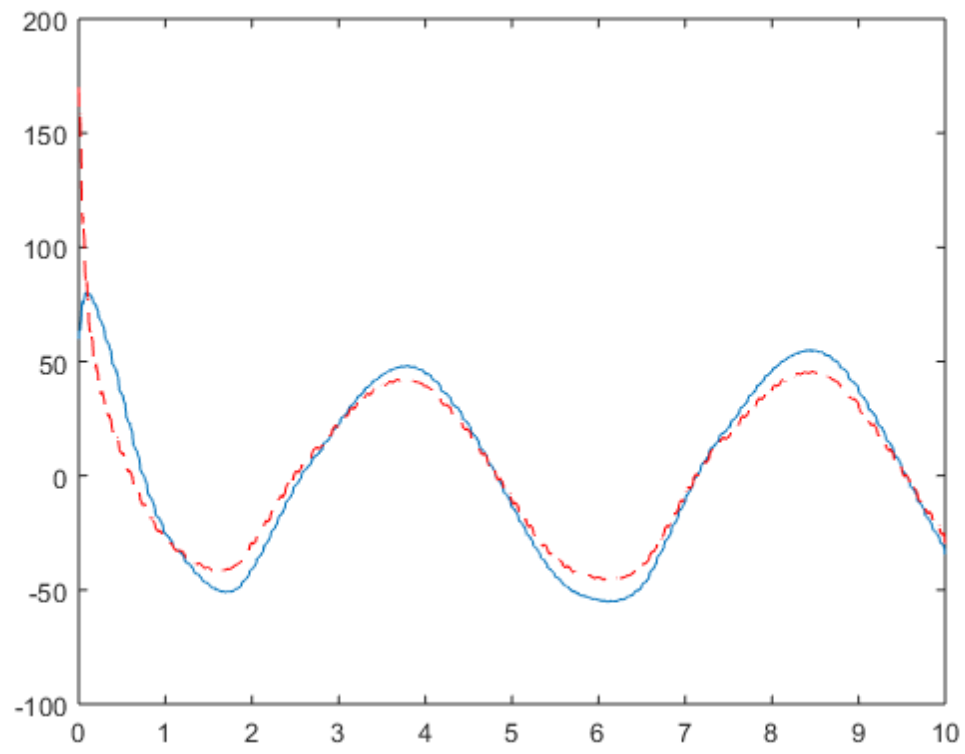
    dx = zeros(4,1);
    dx(1)= x(3); %dtheta1
    dx(2)= x(4); %dtheta2
    dx(3:4) = -invMC* x(3:4) + invM*tau;
end

function tau = Controler(theta_d,dtheta_d,ddtheta_d,theta,dtheta)
    P_e = theta_d - theta;
    V_e = dtheta_d - dtheta;
    Kp = 100*eye(2);
    Kv = 100*eye(2);

    global M C Gq_d
    tau = (Kp*P_e + Kv*V_e) + M*ddtheta_d + C*dtheta_d ;
end

```





## Assignment 4

By- Aishwary Jagetia

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- Iterative Learning control

### Notations: For a given variable, $x$ , $\dot{x}$ is its time derivative, $\ddot{x}$ is

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2nd-order derivative.

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clc
clear all;
close all;
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I1=10; I2 = 10; m1=5; r1=.5; m2=5; r2=.5; l1=1; l2=1; g=9.8;

% we compute the parameters in the dynamic model
a = I1+I2+m1*r1^2+ m2*(l1^2+ r2^2);
b = m2*l1*r2;
d = I2+ m2*r2^2;
```

### create symbolic variable for $x$ .

---

$x_1$  -  $\theta_1$   $x_2$  -  $\theta_2$

```
symx= sym('symx',[4,1]);

global M C Gq dGq
M = [a+2*b*cos(symx(2)), d+b*cos(symx(2));
     d+b*cos(symx(2)), d];
C = [-b*sin(symx(2))*symx(4), -b*sin(symx(2))*(symx(3)+symx(4)); b*sin(symx(2))*symx(3),0];
invM = inv(M);
invMC= inv(M)*C;

% Gravity Matrix
g1=-(m1+m2)*g*l1*sin(symx(2))-m2*g*l2*sin(symx(1)+symx(2));
g2=-m2*g*l2*sin(symx(1)+symx(2));
Gq=[g1;g2];
dGq = [diff(Gq, symx(1)), diff(Gq, symx(2))];
```

### specify your initial and final condition.

---

```
x0= [-0.5,0.2,0.1,0.1];
global i w torque
i = 0;
w = 0.2;
tf=10;
torque = [];
```

### Implement the Iterative Learning control for set point tracking.

---

```

xf = [0, 0, 0, 0];
options = odeset('RelTol',1e-4,'AbsTol',[1e-4, 1e-4, 1e-4, 1e-4]);
[T,X] = ode45(@(t,x) IterativeControl(t,x),[0 tf],x0, options);

figure('Name','Theta_1 under PD SetPoint Control');
plot(T, X(:,1),'r-');
hold on
plot(T, w*ones(size(T,1),1),'b-');
figure('Name','Theta_2 under PD SetPoint Control');
plot(T, X(:,2),'r--');
hold on
plot(T, w*ones(size(T,1),1),'b-');

figure('Name','Torque: Augmented PD Control');
plot(T, torque(1,1:size(T,1)),'-');
hold on
plot(T, torque(2,1:size(T,1)),'r--');
hold off
torque=[];

```

## Iterative Learning control

```

function dx = IterativeControl(t,x)
    global i w
    w=0.2;
    theta_d=[w;w]; % [x1d;x2d] Desired trajectory
    dtheta_d=[0;0]; % [x1d_dot;x2d_dot]
    ddtheta_d=[0;0]; % [x1d_ddot;x2d_ddot]
    theta=x(1:2,1); % [x1;x2]=[x(1);x(2)]
    dtheta=x(3:4,1); % [x1_dot;x2_dot]=[x(3);x(4)]

    global M C Gq dGq
    symx= sym('symx',[4,1]);
    M = subs(M, [symx(1);symx(2);symx(3);symx(4)], [x(1);x(2);x(3);x(4)]);
    C = subs(C, [symx(1);symx(2);symx(3);symx(4)], [x(1);x(2);x(3);x(4)]);
    Gq = subs(Gq, [symx(1);symx(2);symx(3);symx(4)], [x(1);x(2);x(3);x(4)]);
    invM = inv(M);
    invMC= inv(M)*C;
    invMG= inv(M)*Gq;
    dGq = subs(dGq, [symx(1);symx(2)], [x(1);x(2)]);

    tau = Controler(theta_d,dtheta_d,ddtheta_d,theta,dtheta,t);

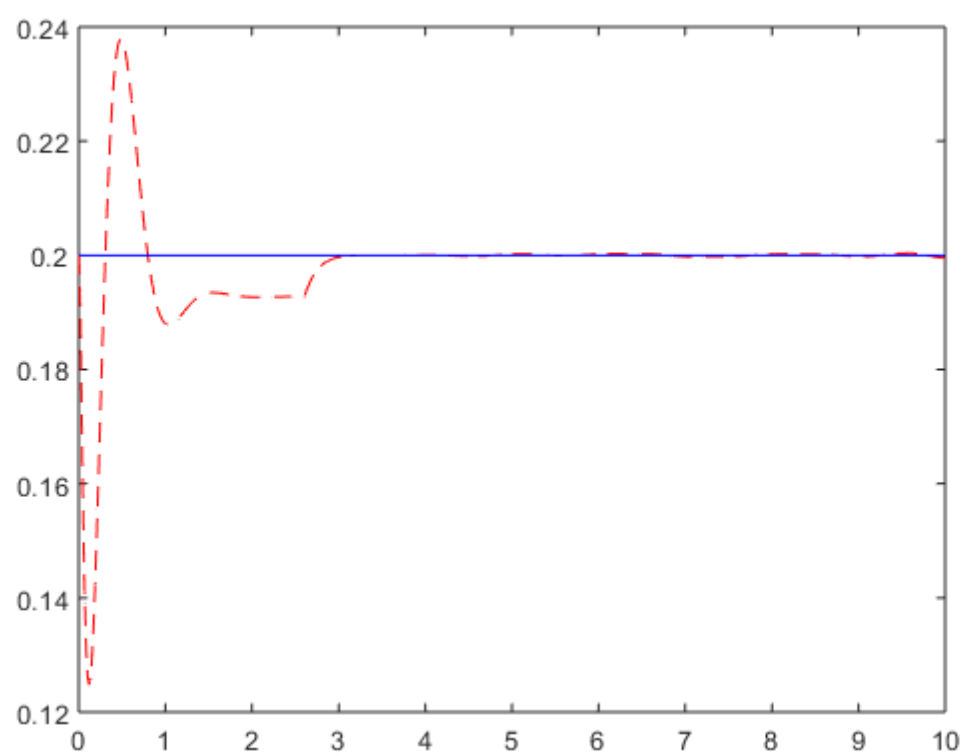
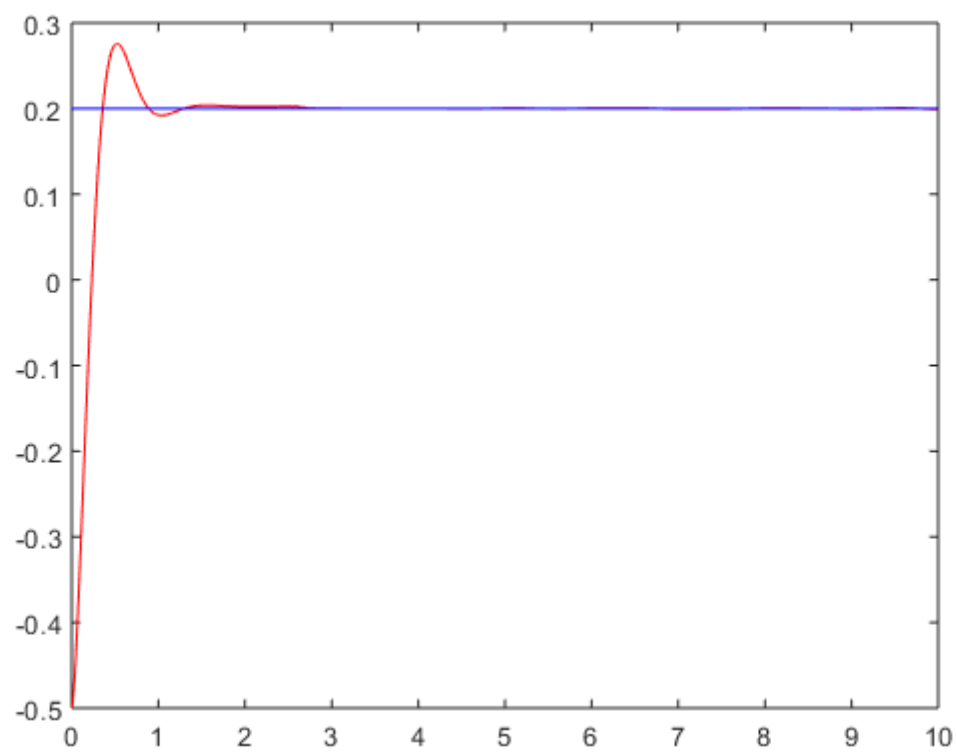
    global torque
    torque = [torque, tau];

    dx = zeros(4,1);
    dx(1)= x(3); %dtheta1
    dx(2)= x(4); %dtheta2
    dx(3:4) = -invMC* x(3:4) + invM*tau -invMG;
end

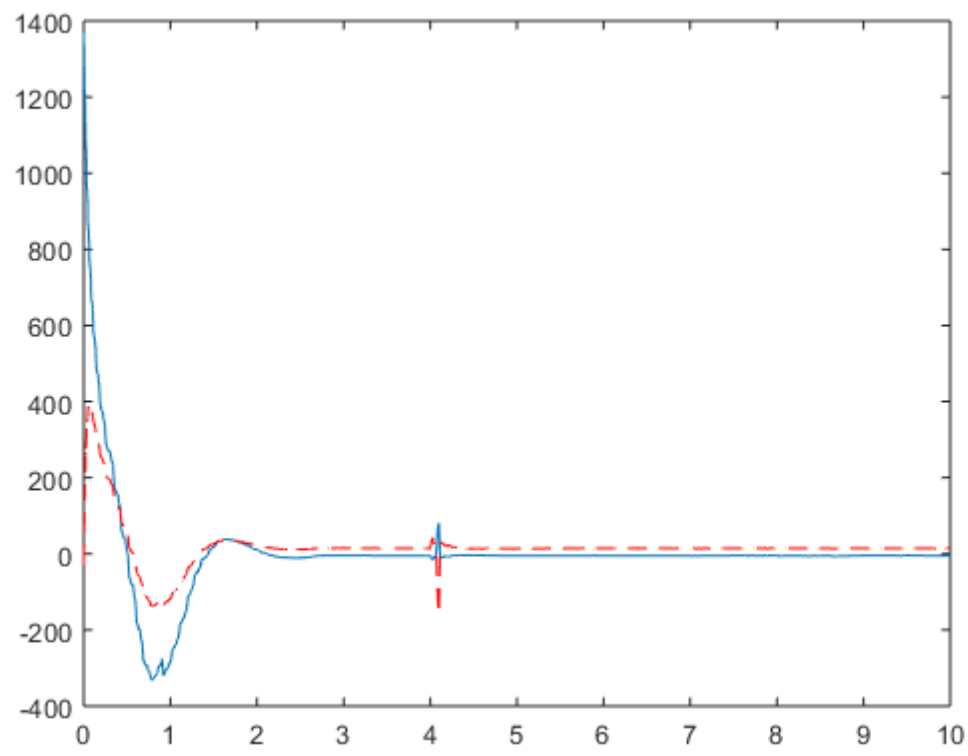
function tau = Controler(theta_d,dtheta_d,ddtheta_d,theta,dtheta,i)
    global u
    P_e = theta_d - theta;
    V_e = dtheta_d - dtheta; % dtheta_d = 0
    % Kp = double(simplify(dGq*eye(2)))
    % To calculate alpha value I have considered calculating partial
    % differentiation of Gq (Gravity Matrix) with respect to joint angles q,
    % which gives me dGq. The max of dGq is used for selecting the value of Kp
    Kp = 200*eye(2);
    Kv = 30*eye(2);
    beta = 0.1;

```

```
if (i == 0)
    u = zeros(2,1);
end
tau = (1/beta)*(Kp*P_e + Kv*V_e) + u;
if (norm(dtheta) < 0.0001)
    u = tau;
end
end
```







#### Assignment #4

For the dynamic model:

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

To find the bound on the gradient of the gravity Term

$$\left\| \frac{\partial N(q)}{\partial q} \right\| \leq \alpha$$

By differentiating gravity matrix w.r.t joint angles

the max value of  $\left\| \frac{\partial N(q)}{\partial q} \right\|$  was coming out to be

around 160, which made me choose  $k_p = 200$  to also satisfy

$$\lambda_{\min}(k_p) \geq \alpha$$

also,

$$\beta = 0.1 \text{ which satisfies } 0 \leq \beta \leq \frac{1}{2}.$$

Thus, the system converges with global asymptotic stability.