

Robot Control Final Exam

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Robust Control:

Step c: Designing Robust Inverse Dynamics control law

- Initially A and B are defined as follows for designing v:

$$A = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

- Control gains k are derived using place function and using eigenvalues $\{-3, -3, -4, -4\}$.
- $A_{cl} = A - B*k$ (A closed loop) is obtained.
- An appropriate positive definite diagonal Q matrix is selected.
- P is obtained using lyap function in matlab.
- A 2X2 diagonal rho matrix is defined.
- Boundary layer phi is selected.
- All these values are used as initial input for ode_robust_RRbot function.

Step d: Updating Ode function

- In the ode function ode_robust_RRbot, all the parameters derived in the above step are initialized.
- Desired states are derived using the equations of the trajectory obtained in the first step.
- Error matrix is defined.
- Then the robust control term vr is obtained using following commands (no boundary layer).

```
if norm(e'*P*B) > 0
    vr = -((e'*P*B)/norm(e'*P*B))*p;
else
    vr = 0;
end
```

- v is obtained as follows.

```
v = vd -k*(e)+vr';
```

- Now the Control inputs are derived as per the dynamics as in feedback linearized controller.

```
T = Mmat*v+Cmat*[dth1; dth2]+Gmat;
```

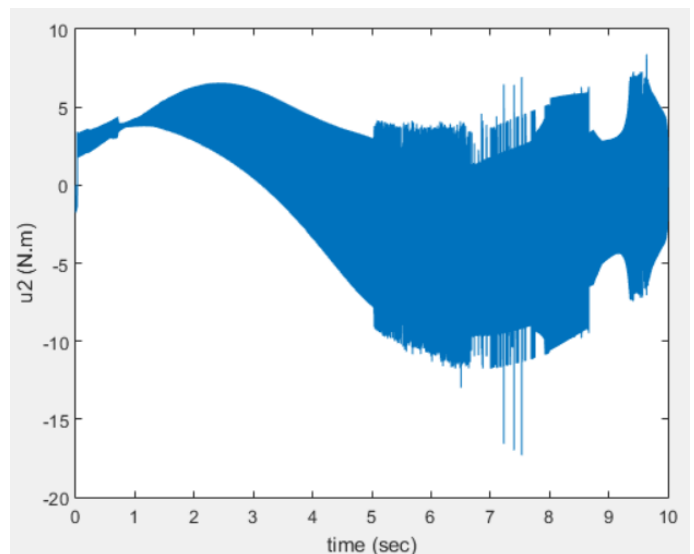
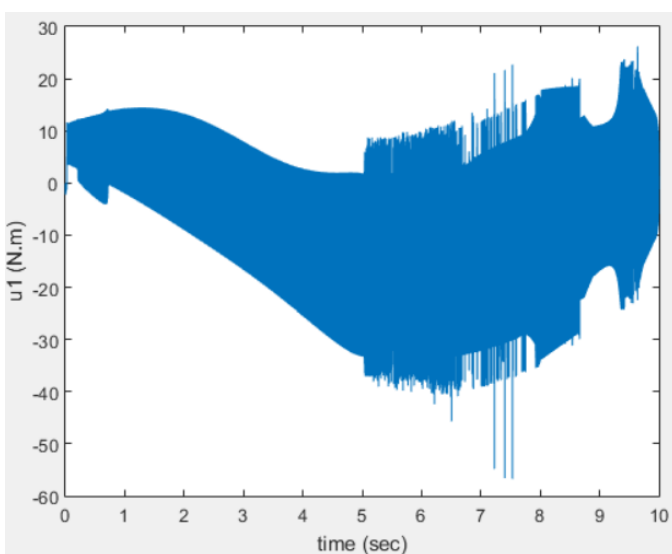
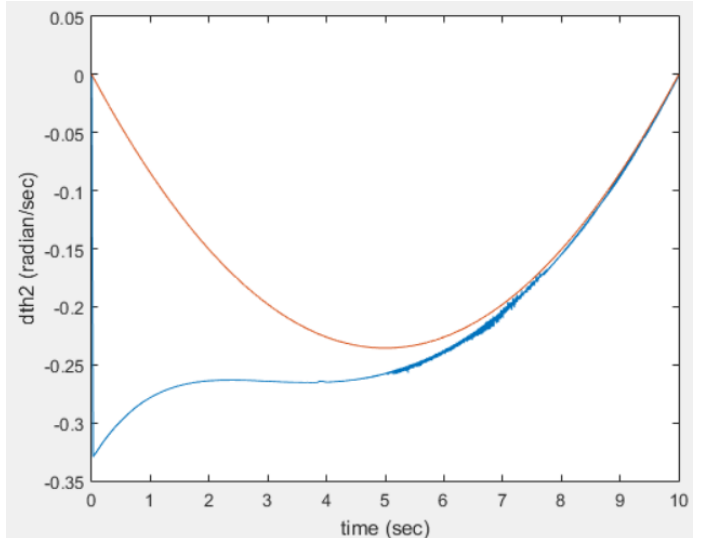
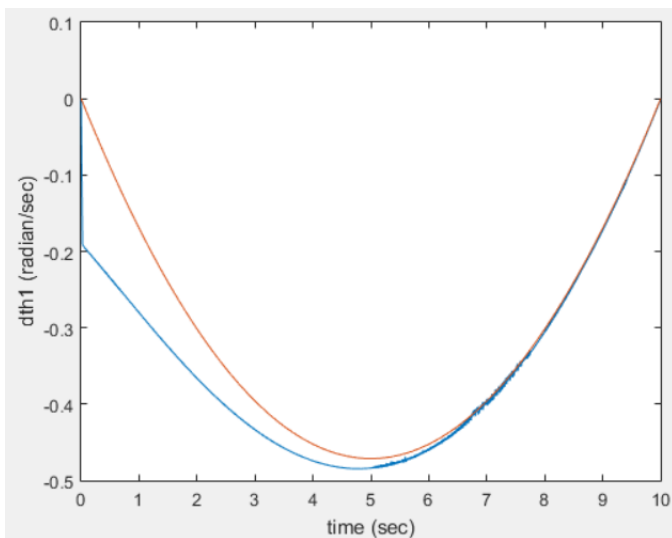
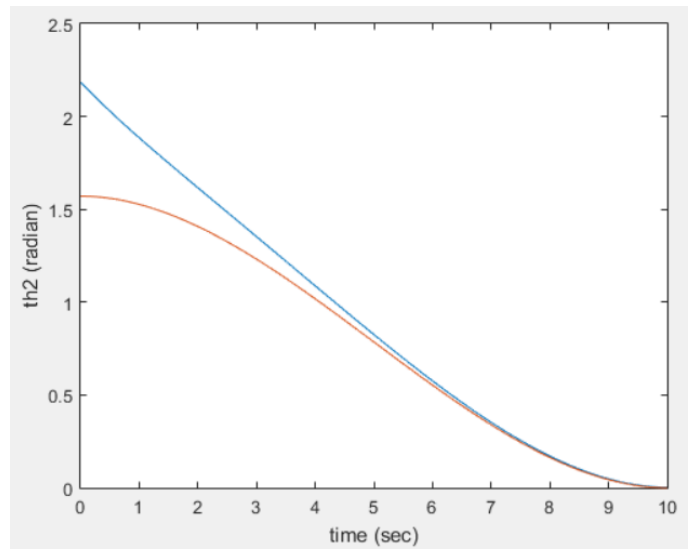
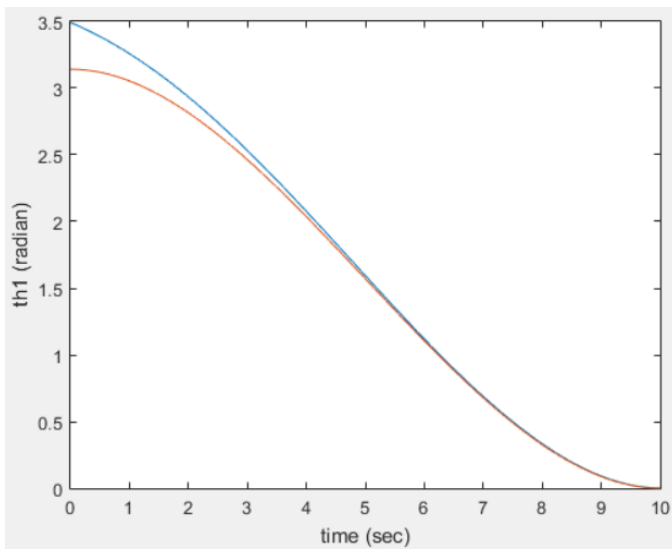
- The findings are simulated using the original values of the dynamics.

Step e: Plotting the output

- List of Parameters after tuning:
 - `lamda = [-3, -3, -4, -4];`
 - `k = place(A,B, lamda);`
 - `Acl = A-B*k;`
 - `Q = eye(4).*5;`
 - `P = lyap(Acl', Q);`

- $p = \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix}; \text{\%rho}$
- $\phi = 0;$

➤ Following outputs were obtained from the simulation.



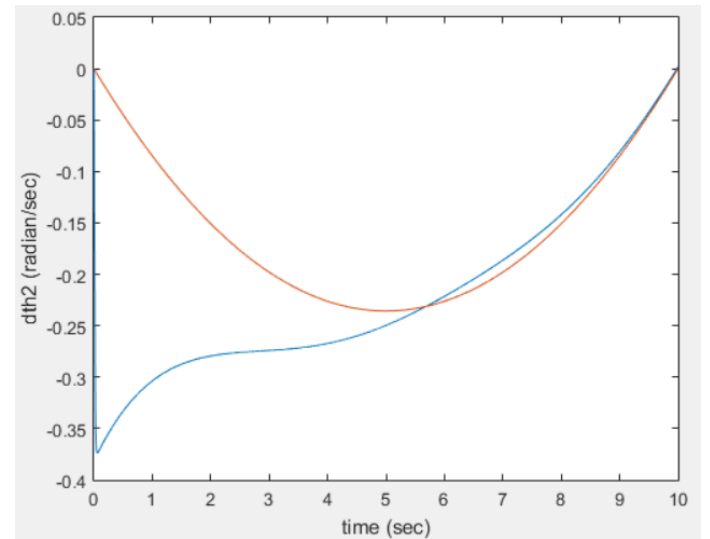
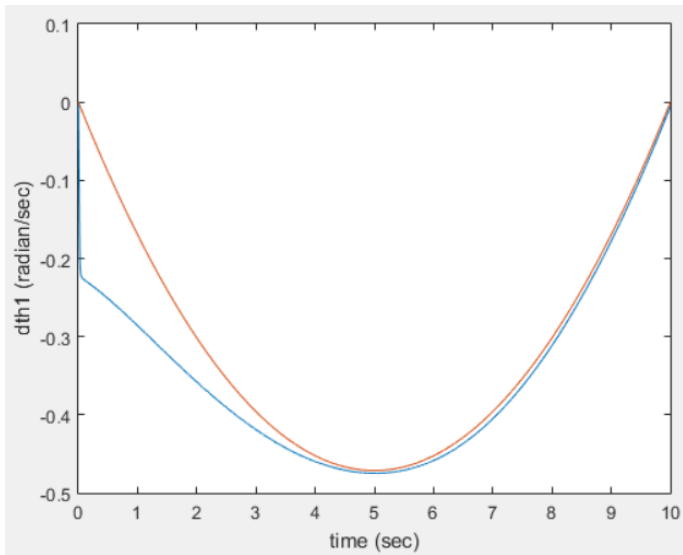
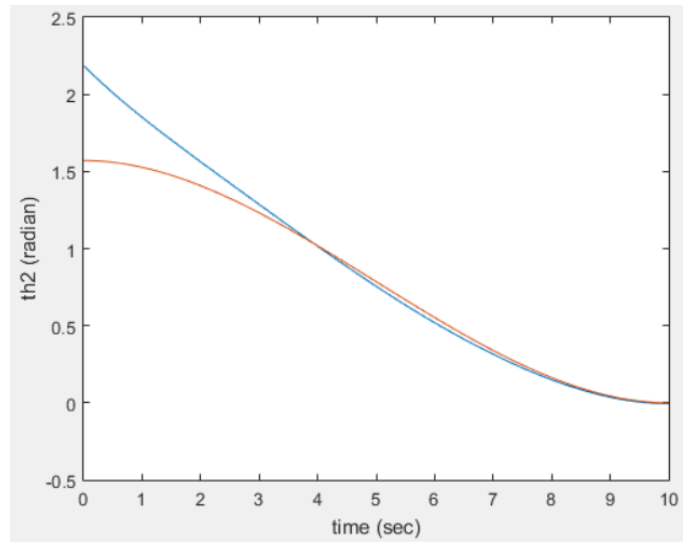
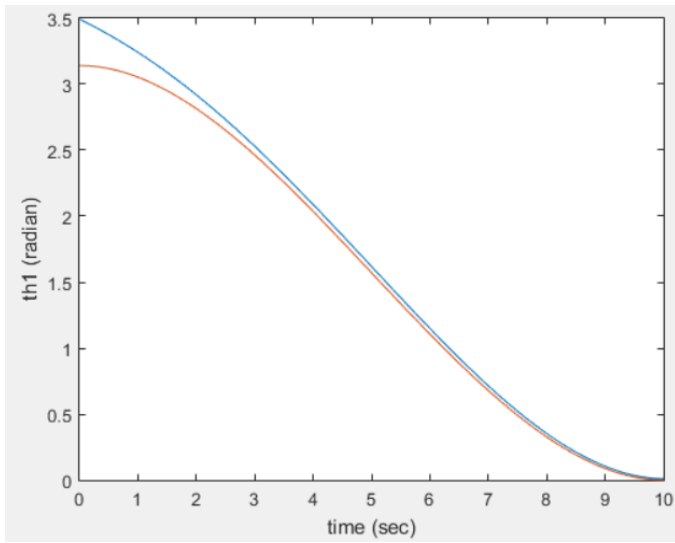
- As seen in the above plots, the system successfully converges to the trajectory.
- But huge amount of chattering was observed in the control efforts and as the control efforts frequently crosses the limits specified in the assignment, the controller frequently fails.
- To overcome this issue, a boundary layer ϕ is defined in the next step.

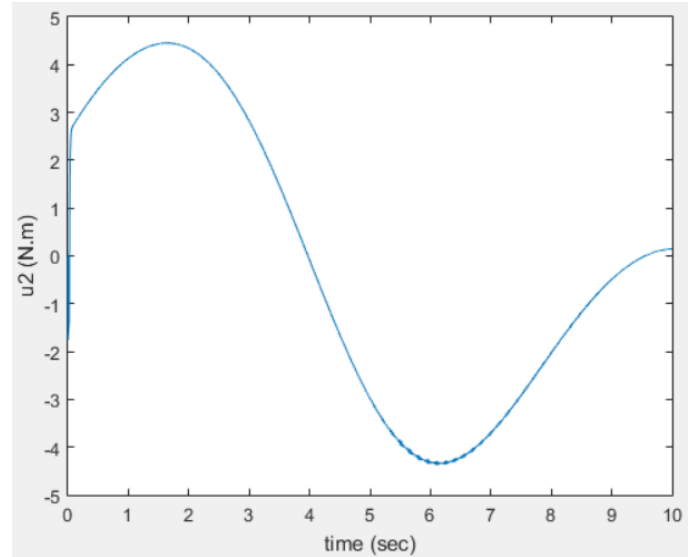
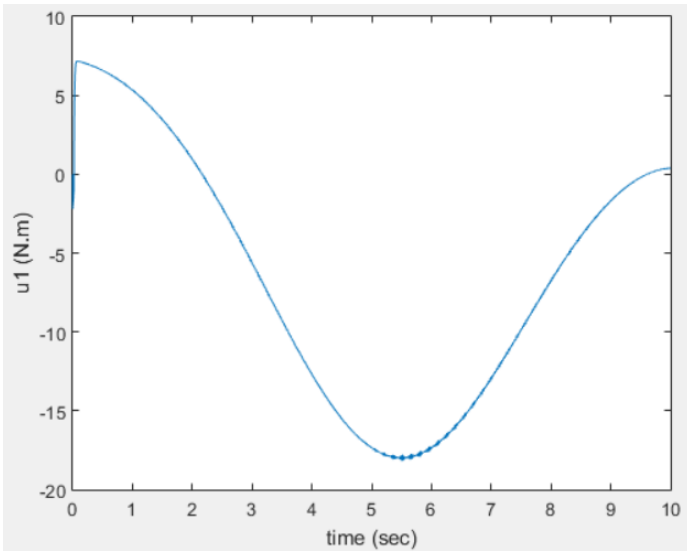
Step f: Defining Boundary layer

- Boundary layer ϕ is defined as follows.

```
if norm(e'*P*B) > phi
    vr = -((e'*P*B)/norm(e'*P*B))*p;
else
    vr = -(e'*P*B)*p/phi;
end
```

- All the parameters were kept unchanged. And $\phi = 0.025$ was implemented.
- Following results were obtained by implementing this boundary layer.

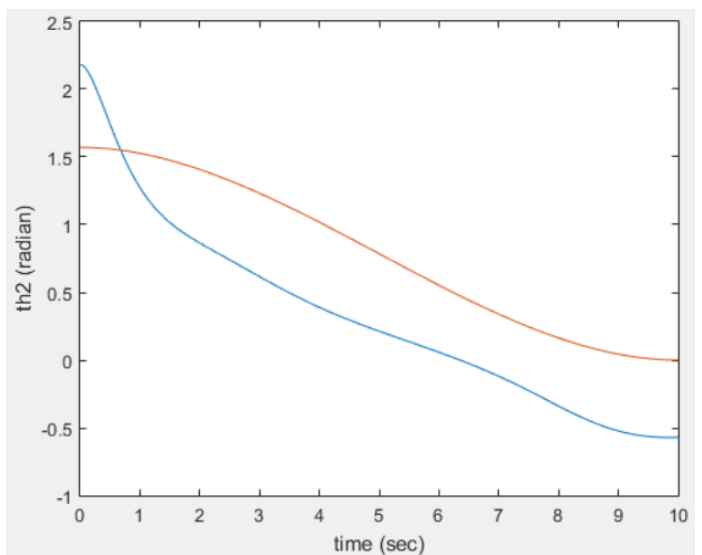
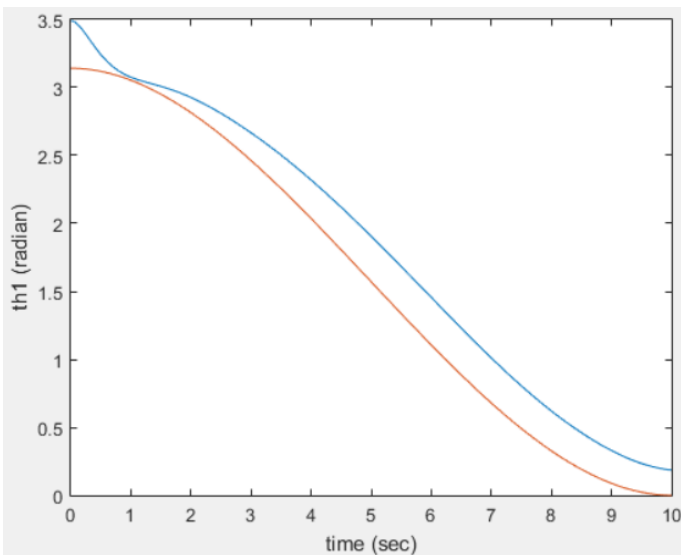


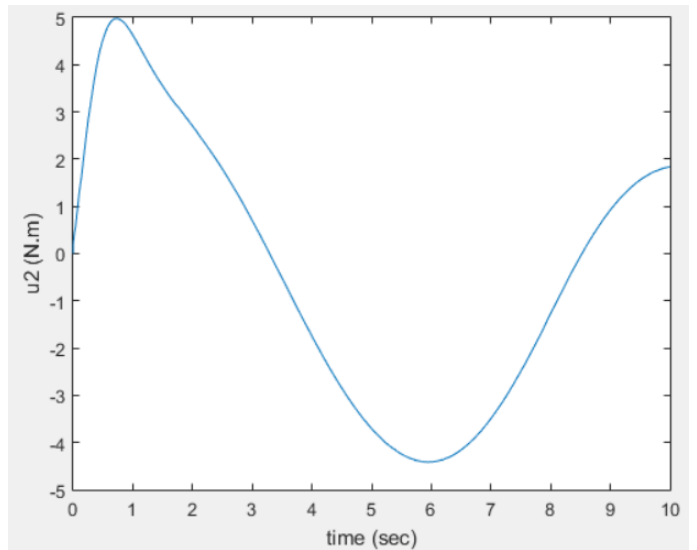
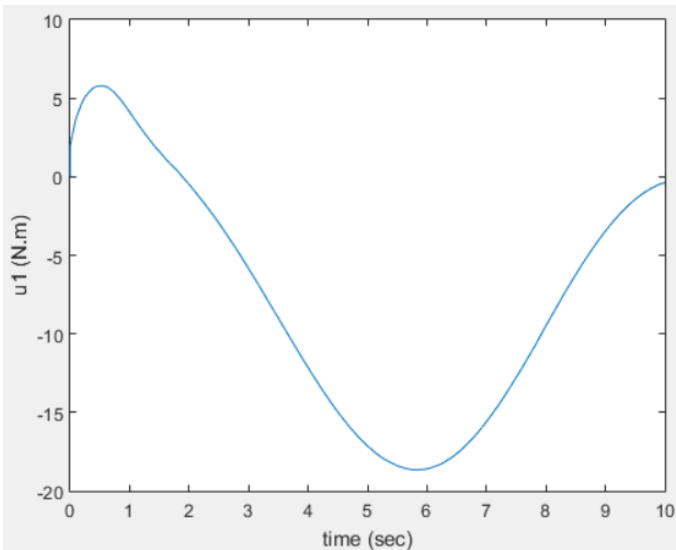
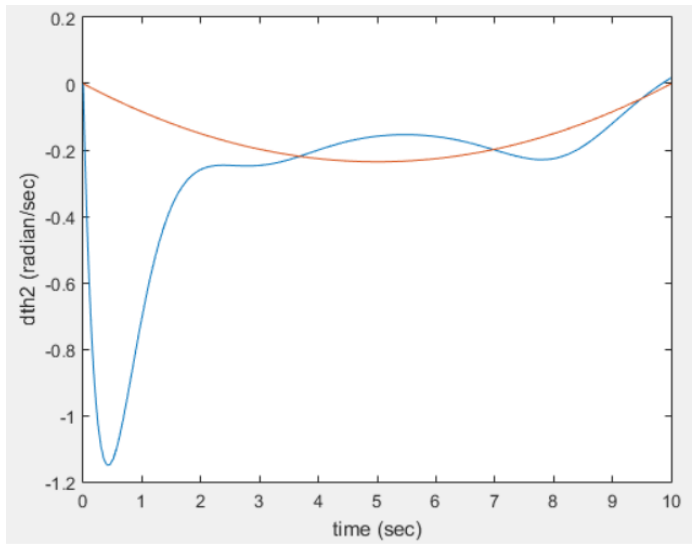
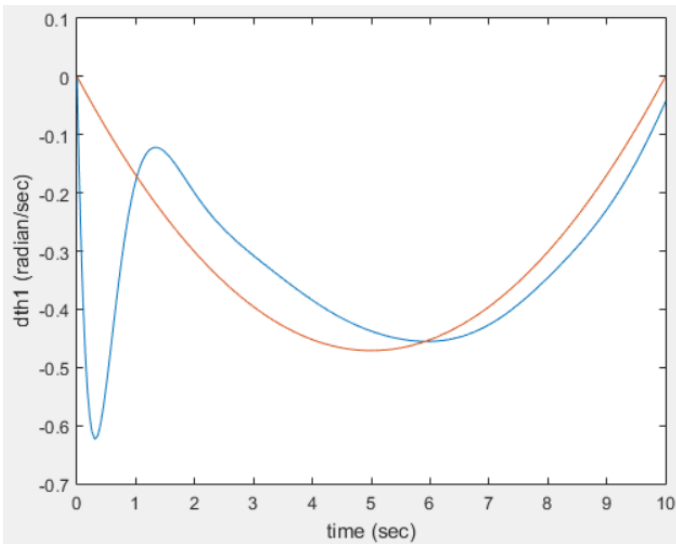


- As seen in the above plots, the system still converges to the trajectory.
- Chattering is greatly reduced.
- The control efforts also falls under the defined range.
- There is a slight error which remains through out the plot which is due to the implementation of the boundary layer.
- The overall output of the controller with boundary layer is more practical then that without boundary layer.

Step g: Comparing with non-robust controller

- Following plots were obtained after setting v_r to zero.

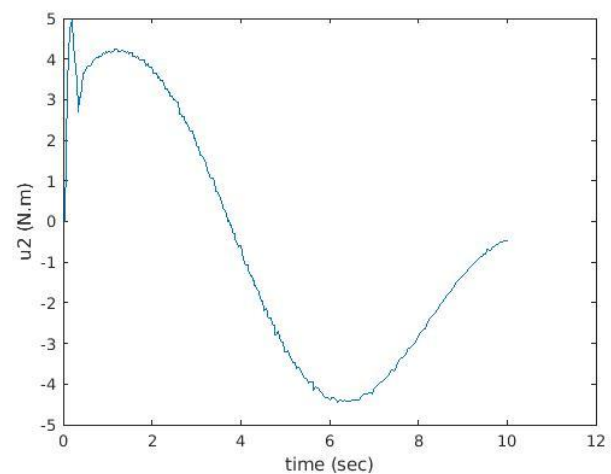
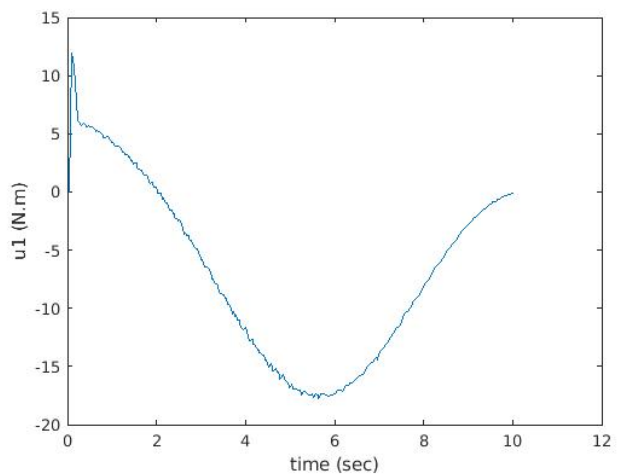
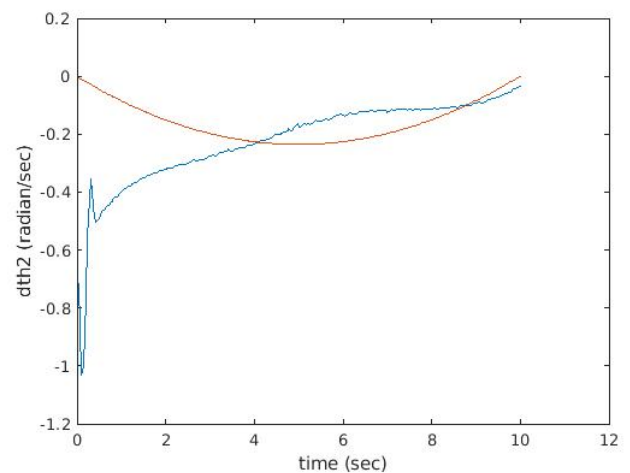
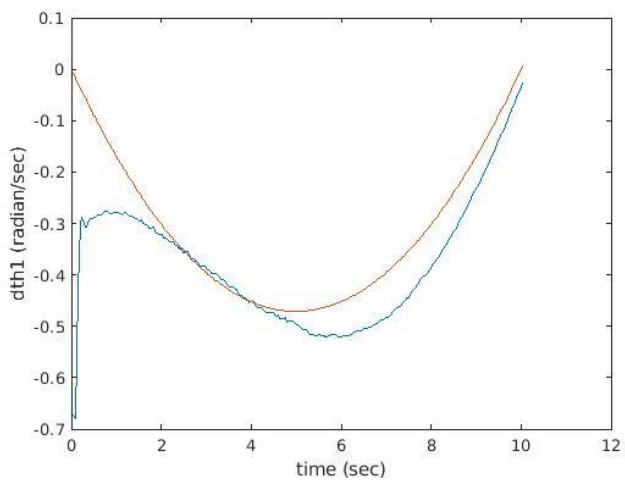
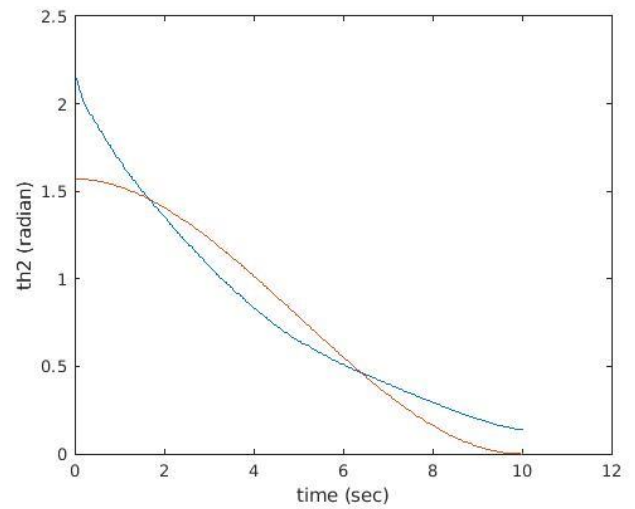
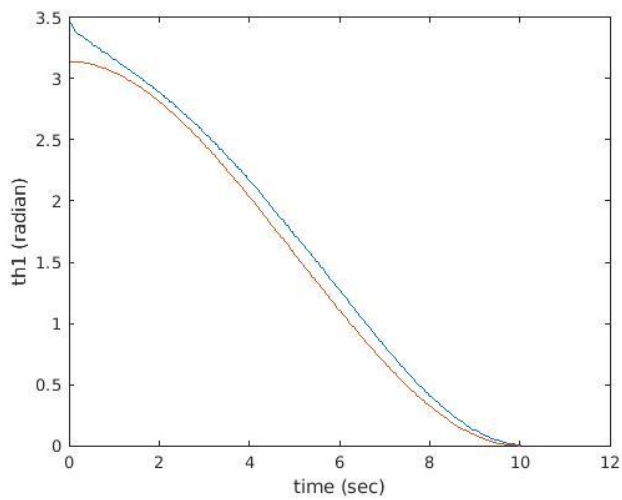




- As seen in the above plots, when v_r is set to zero, the system never converges to the trajectory.
- There is always a steady state error between the system and the desired trajectory.
- It is clearly seen that the robust controller is able to control the system more accurately despite the error in the nominal dynamics.

Step h: Gazebo Implementation

- The same controller discussed above was implemented for the gazebo simulation code `rrobot_robust_control`.
- Initially some chattering were observed but it was removed by tuning the parameters of the robust controller.
- List of parameters after tuning in Gazebo:
 - `lamda = [-3, -3, -4, -4];`
 - `k = place(A,B, lamda);`
 - `Acl = A-B*k;`
 - `Q = eye(4);`
 - `P = lyap(Acl', Q);`
 - `p = [10 0; 0 10]; %rho`
 - `phi = 0.05;`
- Following results were obtained.



- As seen in the plots, the system almost converges to the trajectory.
- To eliminate the slight observed errors, the parameters were increased but it induced the phenomenon of chattering.
- This was the best optimum output obtained after tuning the parameters.
- The control efforts also remains in the defined range.
- It can be seen that the error in between the system and the trajectory is constantly decreasing.

Adaptive Control:

Step j: Designing Adaptive Control Law

- Initially A and B are defined as follows for designing v:

$$A = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

- Control gains k are derived using place function and using eigenvalues $\{-3, -3, -4, -4\}$.
- $A_{cl} = A - B*k$ (A closed loop) is obtained.
- An appropriate positive definite diagonal Q matrix is selected.
- P is obtained using lyap function in matlab.
- Now initial nominal values are defined and the initial value of alpha_hat is obtained.
- Now a 5X5 diagonal matrix gamma is defined.
- All these values are used as initial input for ode_adaptive_RRbot function.

Step k: Updating ode function

- In the ode function ode_adaptive_RRbot, all the parameters derived in the above step are initialized.
- Desired states are derived using the equations of the trajectory obtained in the first step.
- Error matrix is defined.
- Using this error matrix, and desired acceleration, v is calculated.
- A Y_out is defined by replacing all ddth1 and ddth2 with v(1) and v(2) so that it can be directly used to obtain control efforts by multiplying it with alpha_hat.

```
Y_out = [v(1), ...  
         cos(th2)*(2*v(1) + v(2)) - 2*sin(th2)*dth1*dth2 - sin(th2)*dth2^2, ...  
         v(2), ...  
         -sin(th1)*g, ...  
         -sin(th1 + th2)*g; ...  
         0, ...  
         sin(th2)*dth1^2 + cos(th2)*v(1), ...  
         v(1) + v(2), ...  
         0, ...  
         -sin(th1+th2)*g];  
  
T = Y_out*alpha_hat;
```

- This control effort is then substituted in the original dynamics to simulate and also the obtained values are used as a feedback to update the adaption law.
- Now using the obtained values of dth1, dth2, ddth1 and ddth2, Y matrix is obtained.
- M_hat matrix and hence phi matrix are obtained as follows.

```
M1 = [1, 2*cos(th2), 0, 0, 0;  
      0, cos(th2), 1, 0, 0];  
M2 = [0, cos(th2), 1, 0, 0;  
      0, 0, 1, 0, 0];  
M_hat = [M1*alpha_hat, M2*alpha_hat];  
phi = M_hat\Y;
```

- Here M1 and M2 are obtained from Y matrix.
- From these matrices d_alpha_hat is calculated.

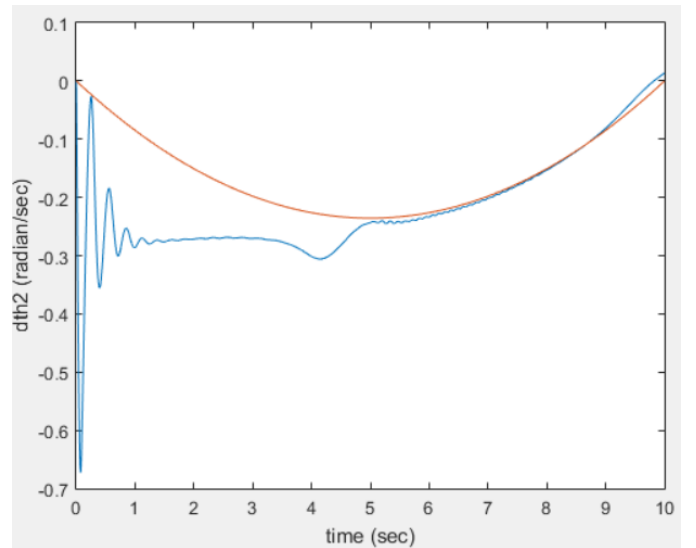
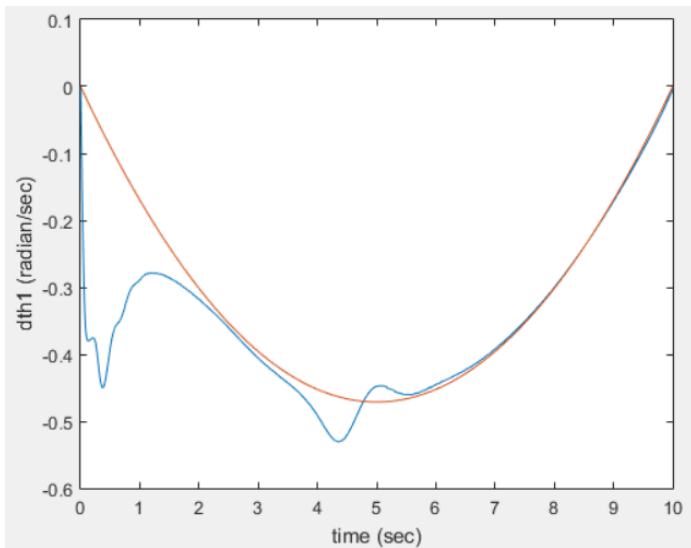
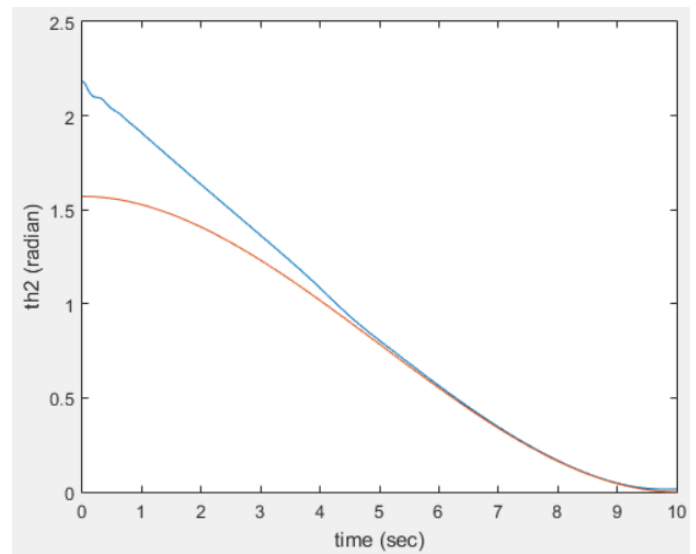
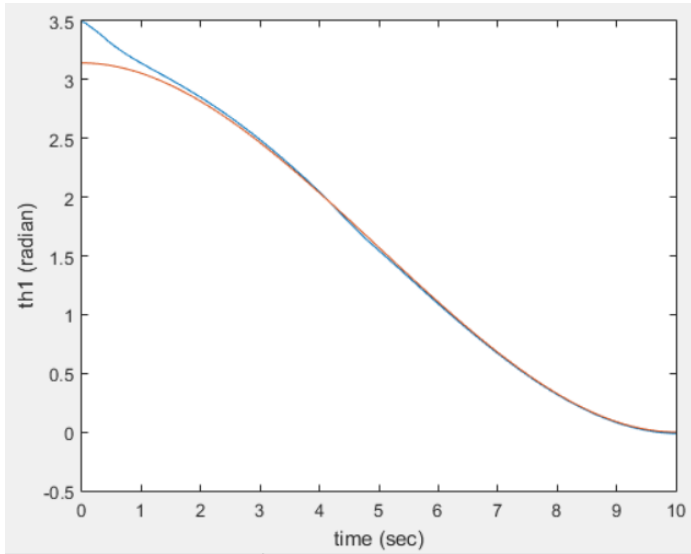
```
d_alpha_hat = -gamma\ (phi'*B'*P*e);
```

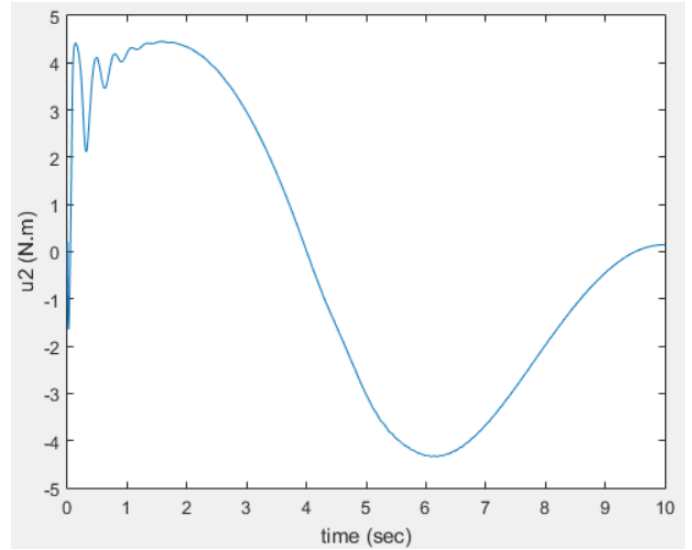
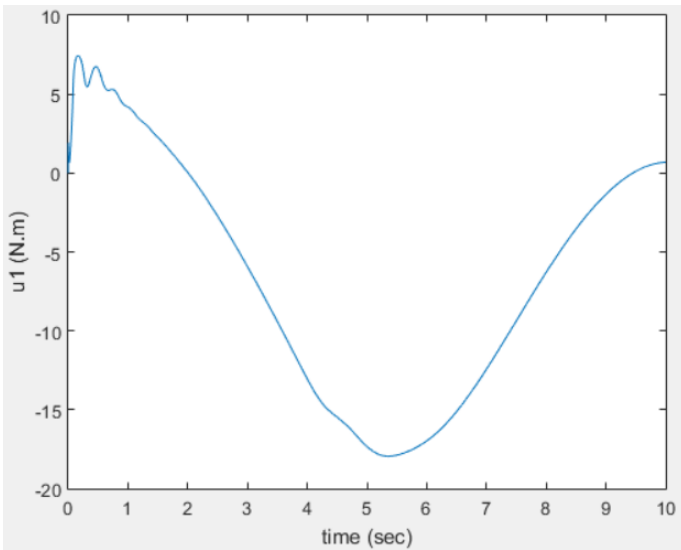
- This is then integrated to update the value of alpha_hat.

- This updated value is now used as updated dynamics for the next iteration.

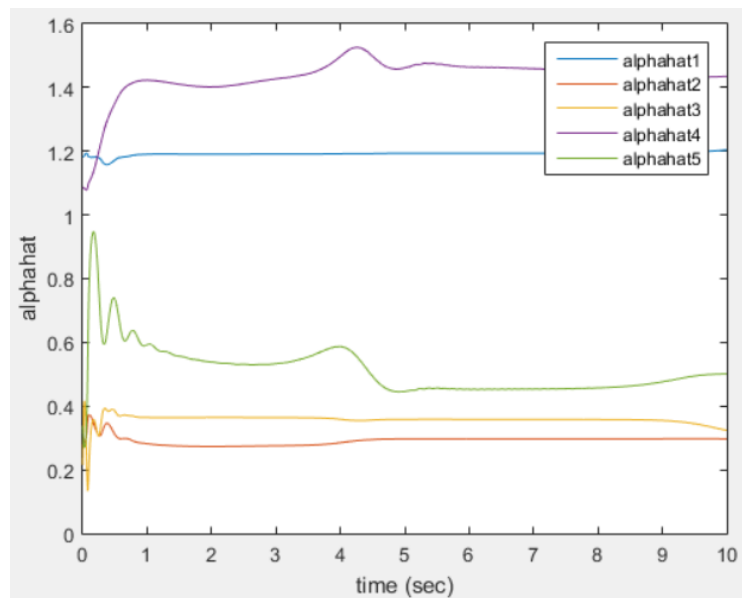
Step 1: Plotting the output

- List of Parameters after tuning:
 - `lamda = [-3, -3, -4, -4];`
 - `k = place(A,B, lamda);`
 - `Ac1 = A-B*k;`
 - `Q = eye(4);`
 - `P = lyap(Ac1', Q);`
 - `gamma = eye(5).*0.1;`
- Following output was obtained after tuning the parameters.





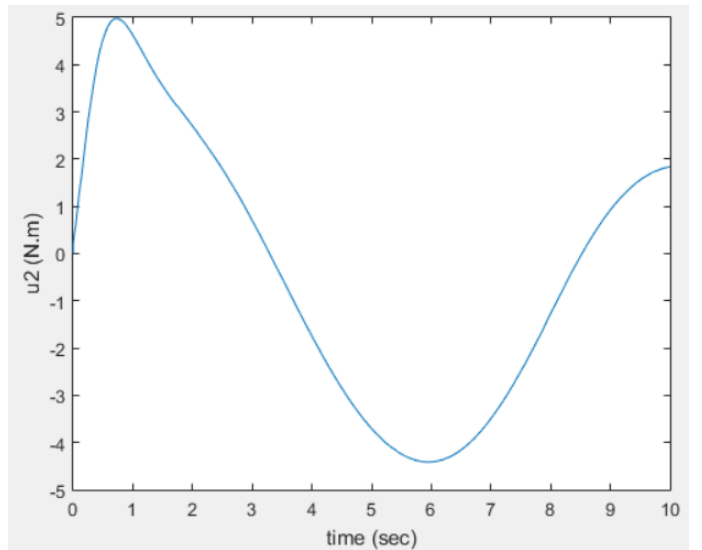
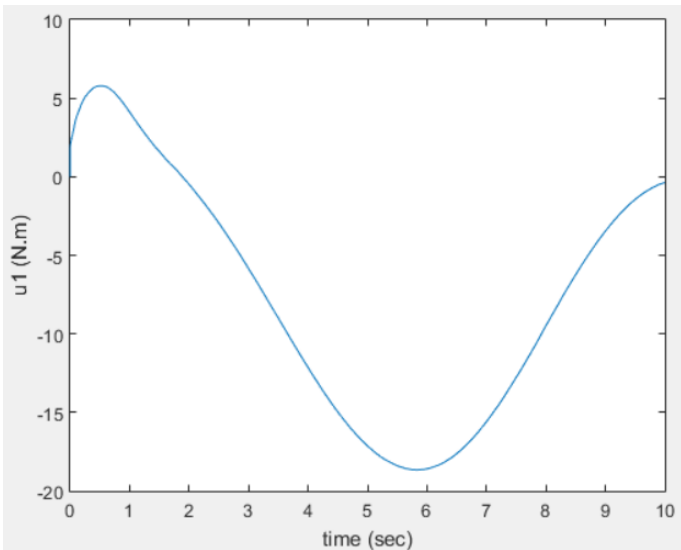
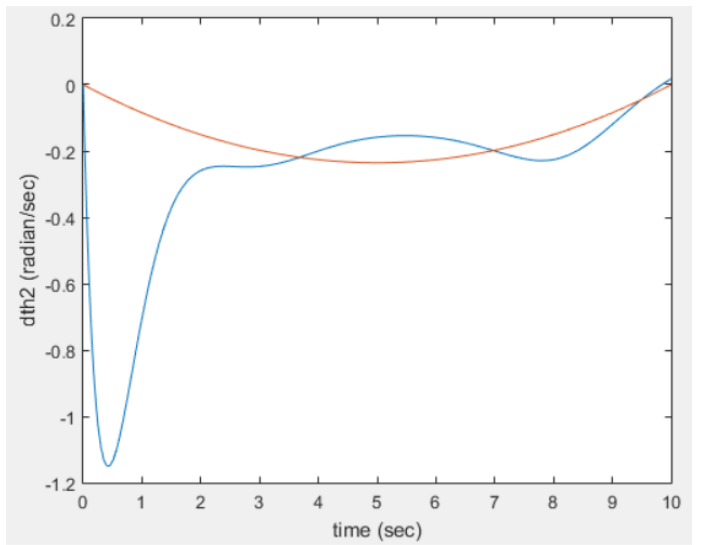
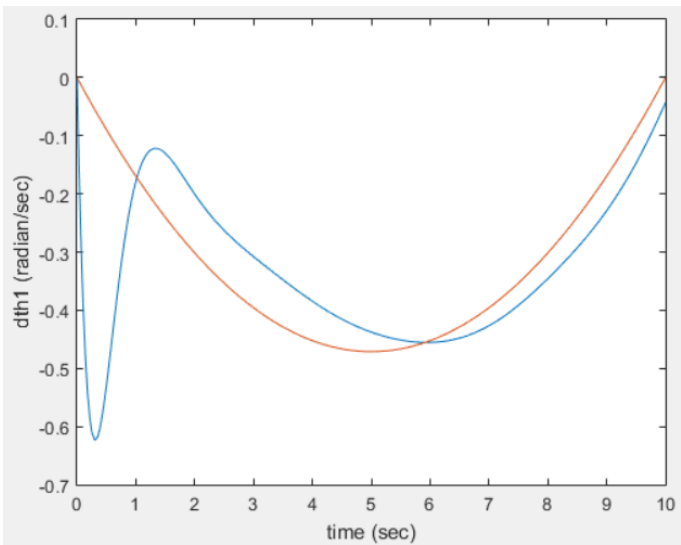
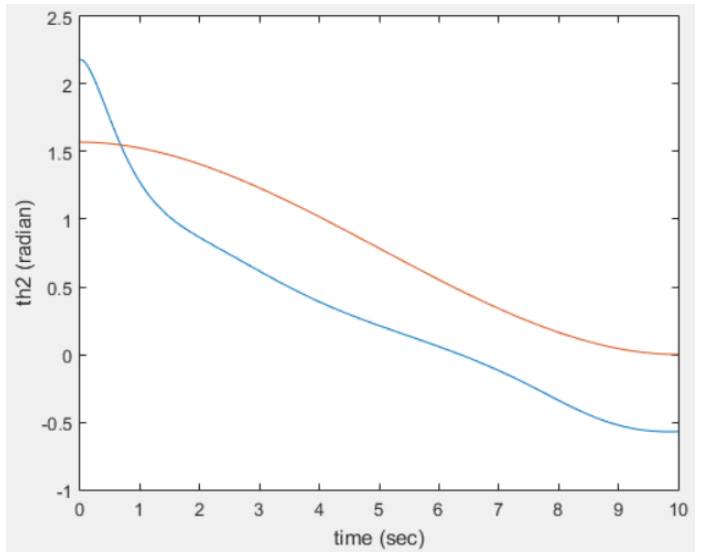
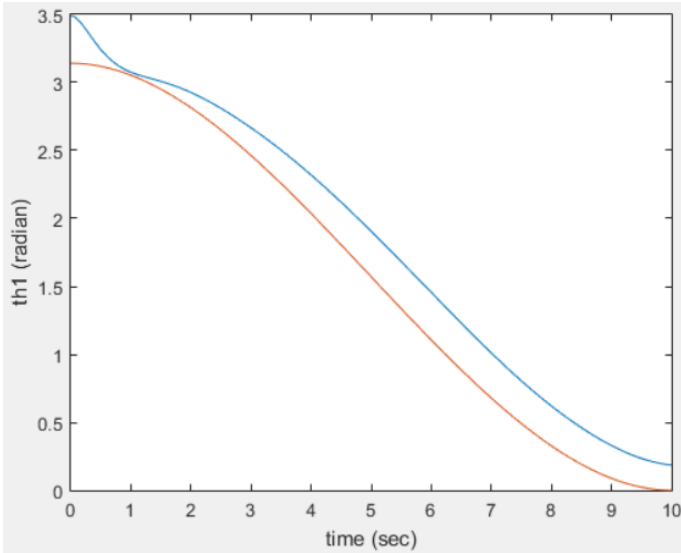
- As seen in the above plots, the system successfully converges to the trajectory.
- And the control efforts also falls under the defined range.
- There were slight oscillations observed initially in $dth1$ and $dth2$ and also in $u1$ and $u2$.
- But the system eventually stabilizes and converges to the trajectory.
- Following are the plots of $\alpha_{\hat{}}$ vs time.



- It can be seen from the plot that the parameters converge to a constant value as time proceeds.
- The original values of α are $\{1.5730, 0.4500, 0.2865, 1.4500, 0.4500\}$.
- It can be observed that the values of $\alpha_{\hat{}}$ do not converge to the original α values but they do converge to constant values.

Step m: Comparing with non-adaptive controller

- Following plots were obtained after setting value of P to zero.



- As seen in the above plots, when P is set to zero, the system never converges to the trajectory.
- There is always a steady state error between the system and the desired trajectory.
- It is clearly seen that the adaptive controller is able to control the system more accurately despite the error in the nominal dynamics.

Step n: Gazebo Implementation

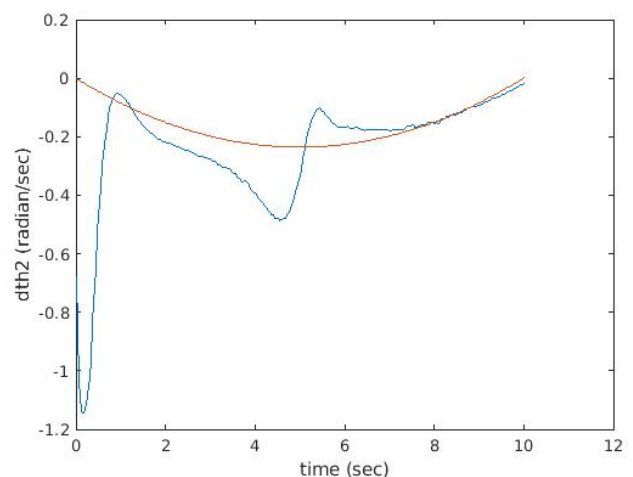
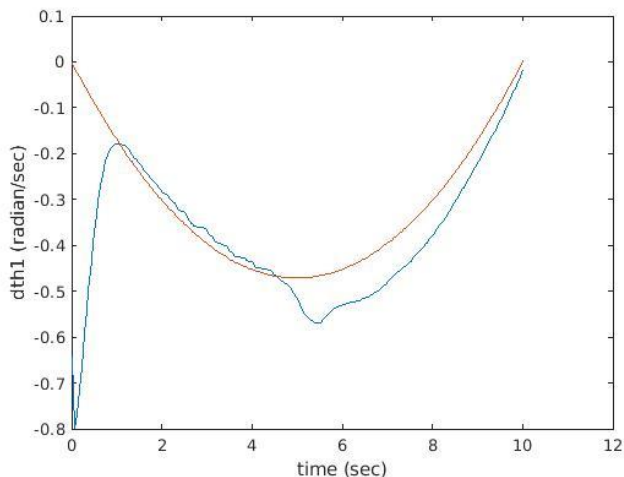
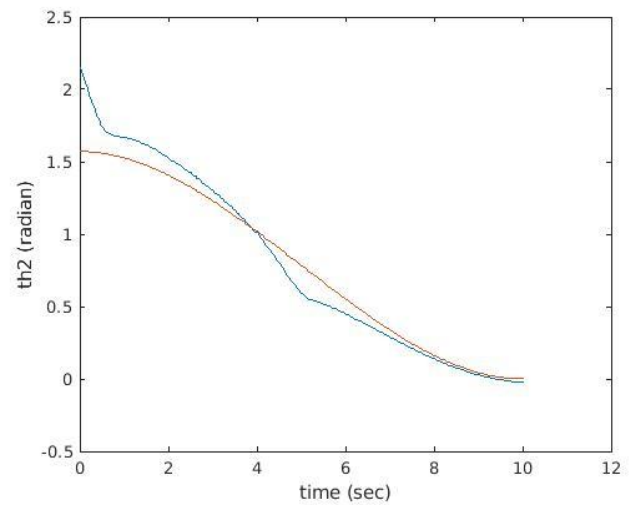
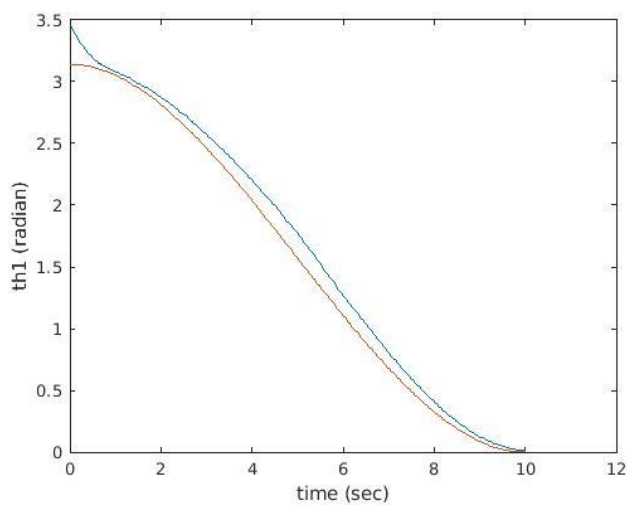
- The same controller discussed above was implemented for the gazebo simulation code `rrobot_adaptive_control`.
- To integrate the $\dot{\alpha}_{\hat{}}$ and find the $\alpha_{\hat{}}$, following iterative formula was used.

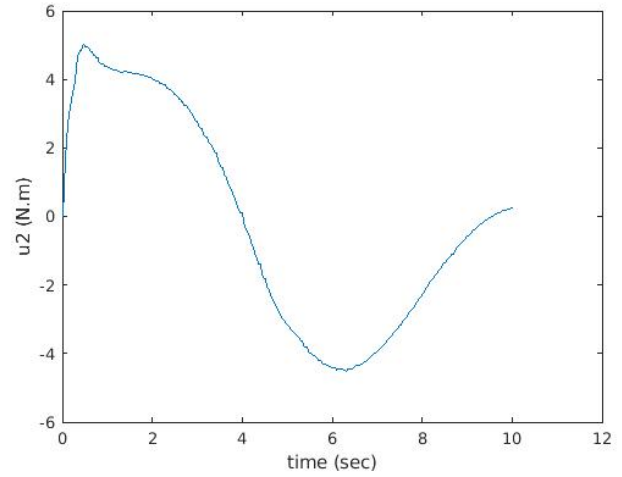
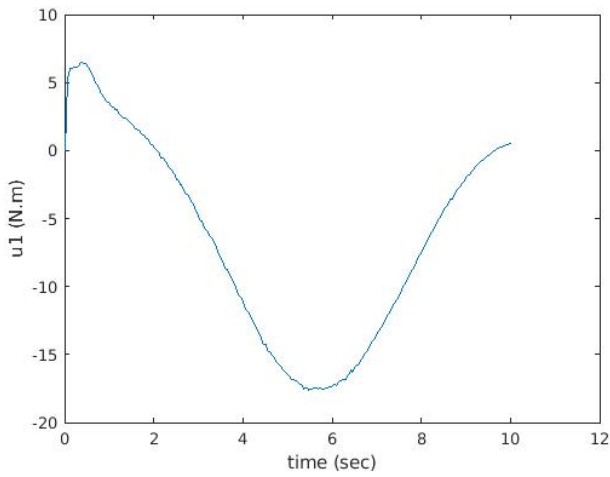
```
alpha_hat = alpha_hat + d_alpha_hat*(t-t_prev);
```

- Other than this ode45 was also used in some simulations to cross verify the results.
- ode45 was implemented as follows.

```
[tm,y] = ode45(@(tm,y) d_alpha_hat, [t_prev t], alpha_hat);
b = size(y);
alpha_hat = y(b(1), :);
```

- There was no difference in the results observed by implementing above two formulas separately.
- But it was observed that using the ode45 was slight computationally heavy as it did around 50 iterations each time and gave a list of values, from which only last value was used.
- List of parameters after tuning in Gazebo:
 - $\lambda = [-3, -3, -4, -4];$
 - $k = \text{place}(A,B,\lambda);$
 - $A_{cl} = A-B*k;$
 - $Q = \text{eye}(4);$
 - $P = \text{lyap}(A_{cl}', Q);$
 - $\gamma = \text{eye}(5).*9;$
- After critically tuning the parameters, following result were obtained.





- As seen in the plots, the system successfully converges to the trajectory.
- The control efforts also remains in the defined range.
- It can be seen that the error in between the system and the trajectory is constantly decreasing.
- It also can be observed from the plots that adaptive controller performed overall better than the robust controller in the gazebo environment.
- Adaptive controller was able to converge system to the trajectory more accurately than the robust controller without any chattering.
- But the parameters of Adaptive controller were very critically tuned, even a small change in the parameters can make the system to explode.