Powertrain Control Final Aravind Chandradoss

1) Observability

The State Space representation of 4-cylinder SI engine is given by,

$$x_{1}(k) = \frac{c_{T2}}{\sqrt{\tau_{o}}} p(k-1) + \frac{c_{T3}}{\sqrt{\tau_{o}}} w(k-1) - \frac{c_{T4}}{\sqrt{\tau_{o}}} \alpha(k-1)$$

$$x_{2}(k) = \frac{c_{T6}}{\sqrt{\tau_{o}}} w(k-1) - \frac{c_{T7}}{\sqrt{\tau_{o}}} T(k-1)$$

$$x_{3}(k) = p(k-1)$$

$$x_{4}(k) = x_{3}(k-1) = p(k-2)$$

$$x_{5}(k) = x_{4}(k-1) = p(k-3)$$

State Space Matrices:

$$x(k+1) = \Phi_{T}x(k) + \Gamma_{T}\alpha(k)$$

$$y(k) = H_{T}x(k) + D_{T}\alpha(k)$$

$$\Phi_{T} = \begin{bmatrix} (OM^{-1})_{11} & (OM^{-1})_{12} & 0 & 0 & Q_{12}K_{p} \\ (OM^{-1})_{21} & (OM^{-1})_{22} & 0 & 0 & Q_{22}K_{p} \\ (M^{-1})_{11} & (M^{-1})_{12} & 0 & 0 & -(M^{-1}N)_{12}K_{p} \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix} \Gamma_{T} = \begin{bmatrix} Q_{11} \\ Q_{21} \\ -(M^{-1}N)_{11} \\ 0 \\ 0 \end{bmatrix}$$

$$H_{T} = \begin{bmatrix} (M^{-1})_{21} & (M^{-1})_{22} & 0 & 0 & -(M^{-1}N)_{22}K_{p} \end{bmatrix} \quad D_{T} = \begin{bmatrix} -(M^{-1}N)_{21} \end{bmatrix}$$

$$M = \begin{bmatrix} \frac{-C_{T_{1}}}{\sqrt{T_{0}}} & \frac{-C_{T_{3}}}{\sqrt{T_{0}}} \\ 0 & \frac{-C_{T_{5}}}{\sqrt{T_{5}}} \end{bmatrix}$$

$$N = \begin{bmatrix} \frac{C_{T_{4}}}{\sqrt{T_{0}}} & 0 \\ 0 & \frac{C_{T_{7}}}{\sqrt{T_{5}}} \end{bmatrix}$$

$$O = \begin{bmatrix} \frac{C_{T_2}}{\sqrt{\tau_0}} & \frac{C_{T_3}}{\sqrt{\tau_0}} \\ 0 & \frac{C_{T_6}}{\sqrt{\tau_0}} \end{bmatrix}$$

$$P = \begin{bmatrix} \frac{-C_{T_4}}{\sqrt{\tau_0}} & 0 \\ 0 & \frac{-C_{T_7}}{\sqrt{\tau_0}} \end{bmatrix}$$

$$Q = -OM^{-1}N + P$$

In general, X_{ij} stands for the element in $i^{th}row$ and $j^{th}column$

$$\Phi_T = \begin{bmatrix} 0.9724 & -6.1276 & 0 & 0 & 0.0011 \\ 0 & 0.9990 & 0 & 0 & -0.0004 \\ -1.0092 & 3.1353 & 0 & 0 & -0.0006 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix} \qquad \Gamma_T = \begin{bmatrix} -91.2330 \\ 0 \\ 46.6806 \\ 0 \\ 0 \end{bmatrix}$$

$$H_T = \begin{bmatrix} 0 & -1.0228 & 0 & 0 & 0.0002 \end{bmatrix} \qquad D_T = \begin{bmatrix} 0 \end{bmatrix}$$

Observability:

Rank of observability matrix is 5 (=n). Thus, the system is observable.

$$Rank\left(W_{o}\right)=Rank(\begin{bmatrix}\mathbf{H}_{t} & \mathbf{H}_{t}\phi_{t} & \mathbf{H}_{t}\phi_{t}^{2} & \mathbf{H}_{t}\phi_{t}^{3} & \mathbf{H}_{t}\phi_{t}^{4}\end{bmatrix}^{T}\Big)=5=n\,,$$

2. Observer design

The observer is designed such that,

$$u(k) = -K x_a(k)$$

Where, $x_a(k) = [x(k) \ x_I(k)]$. Once the state are estimated, the estimated states will be used for state feedback.

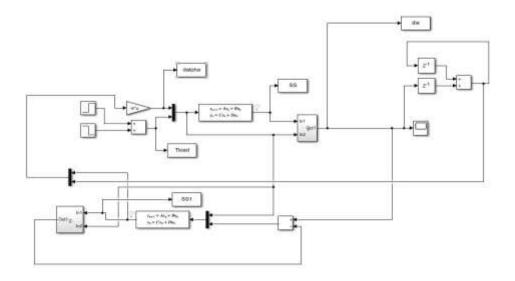
$$u(k) = -K x^{\wedge}_{a}(k)$$

Where,
$$x^{*}_{a}(k) = [x^{*}(k) \ x_{I}(k)].$$

The state feedback gain for Luenberger gain is found using pole placement. The poles were places more towards the origin to have fast response. The chosen poles were as follows,

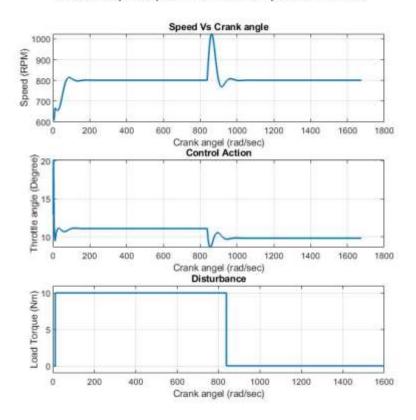
The feedback gain obtained using pole placement is

As expected the gain were high to have fast convergence.



The response for state feedback using estimated state is as follows,

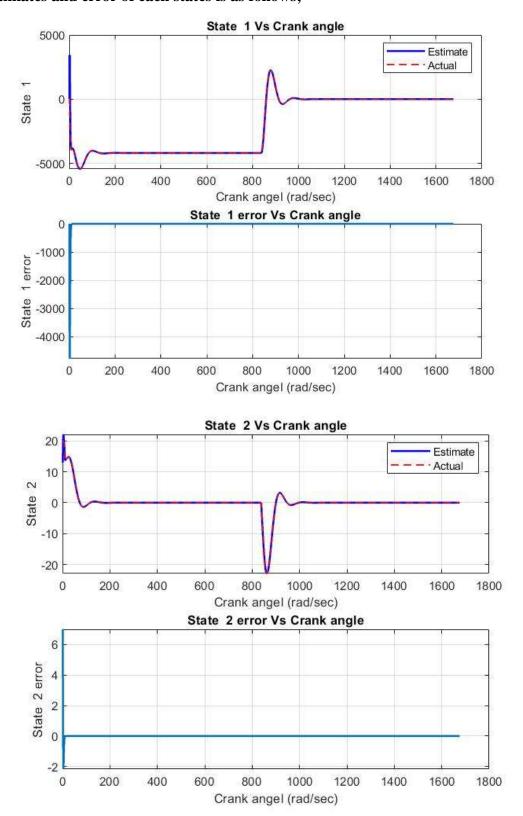
Closedloop Response for load torque disturbance

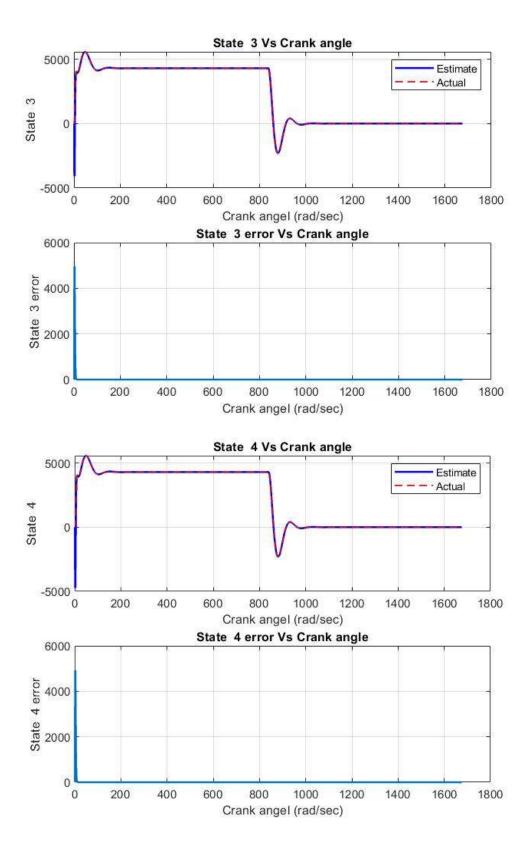


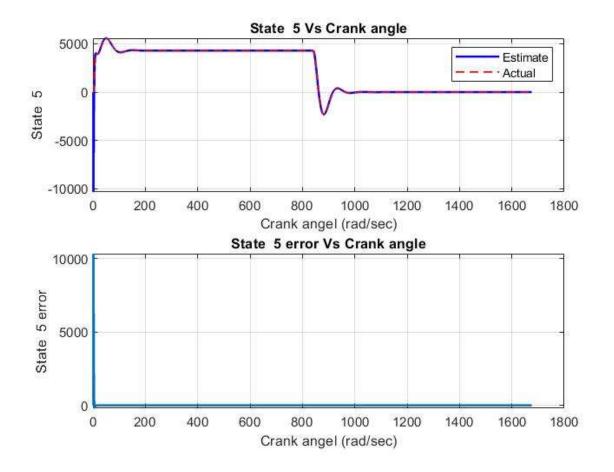
Closed loop Response using estimated state

NOTE: The speed in rpm starts at approx. 650rpm because the initial states is chosen non-zero for the plant also (to check regulatory response) and for the observer (to check convergence of the estimates) (both the initial conditions are not equal)

The estimates and error of each states is as follows;







Inference:

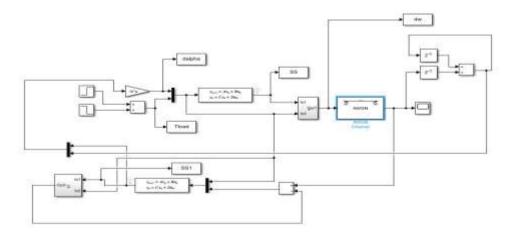
All the state are tracked properly (Error between estimate and actual states tends to zero)

The throttle angle, overshoot, settling time are within desired region.

The steady state error is zero (Because of integral action)

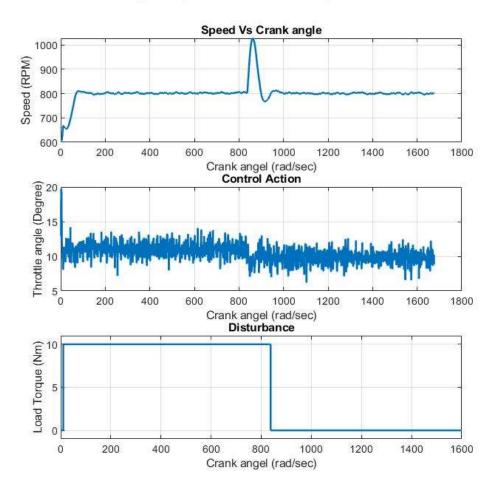
3. Gaussian Noise

Adding Gaussian noise (Variance=0.8) to speed measurement and using the same control as in previous problem.



We get the following response.

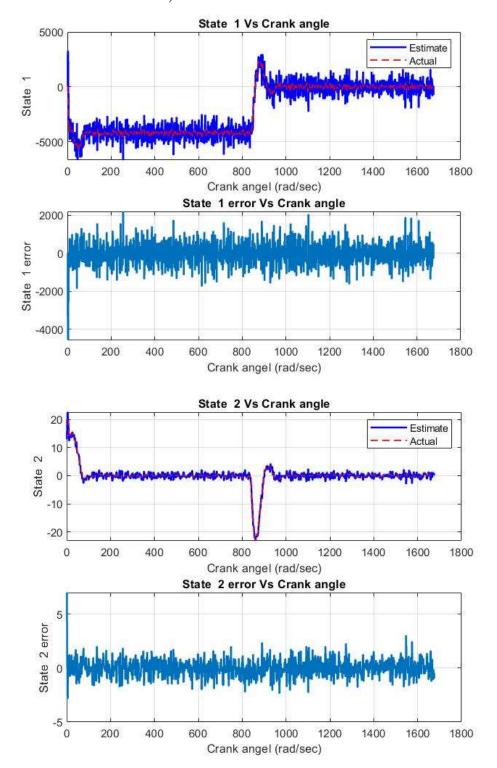
Closedloop Response for load torque disturbance

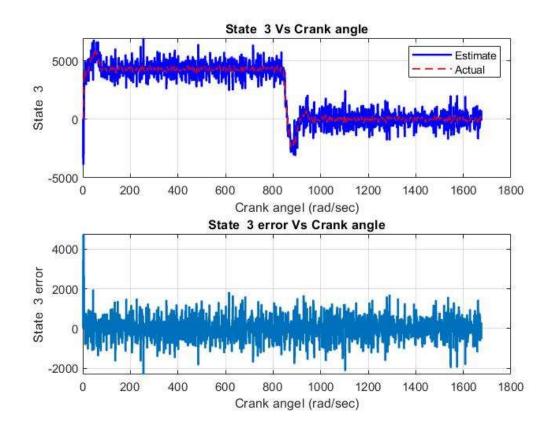


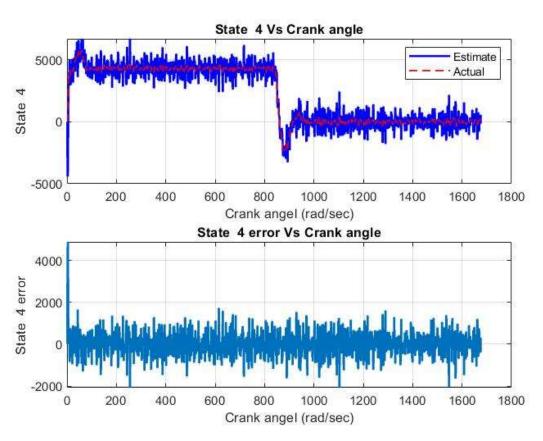
Closed Loop Response with Gaussian noise

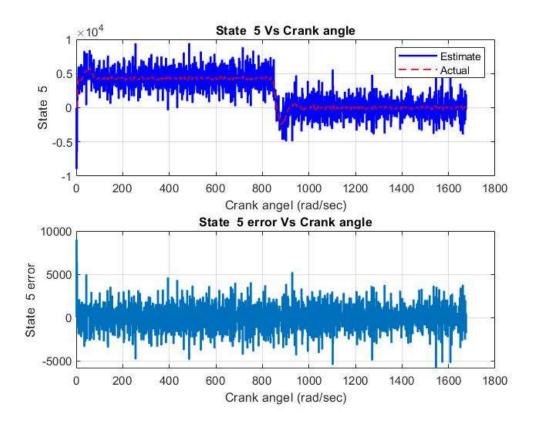
NOTE: The speed in rpm starts at approx. 650rpm because the initial states is chosen non-zero for the plant also (to check regulatory response) and for the observer (to check convergence of the estimates) (both the initial conditions are not equal)

The state estimate and errors are,









Inference:

All the state were tracked significantly (In particular, the states 1 and 2 are tracked with comparably lesser error than the states 3, 4 and 5. This is due to the added Gaussian noise and inherent delay in the system)

The throttle angle, overshoot, settling time are within desired region (But oscillates). The steady state error is non-zero (But chatters around zero value)

4. Kalman Filter design

To the same system (with Gaussian noise), Kalman filter is implemented as follows;

The Process noise (Q) and measurement noise (R) is assumed to be 0.01 and 0.8 respectively. The filter gain is determined using state-state kalman filter gain,

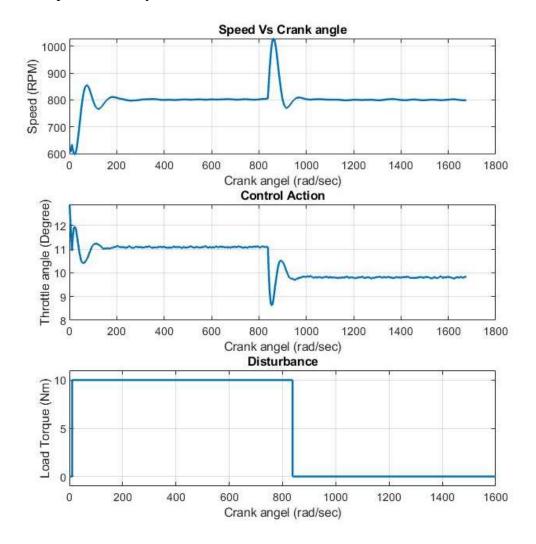
[KEST,K_bar,P]=kalman(Plant,Q_noise,R_noise)

The K_bar is used as filter gain. (With same LQR controller as in previous problem) The system response is as follows,

The kalman filter gain (K_bar) is,

0.4340 -0.0076 -0.4259 -0.3892 -0.3524

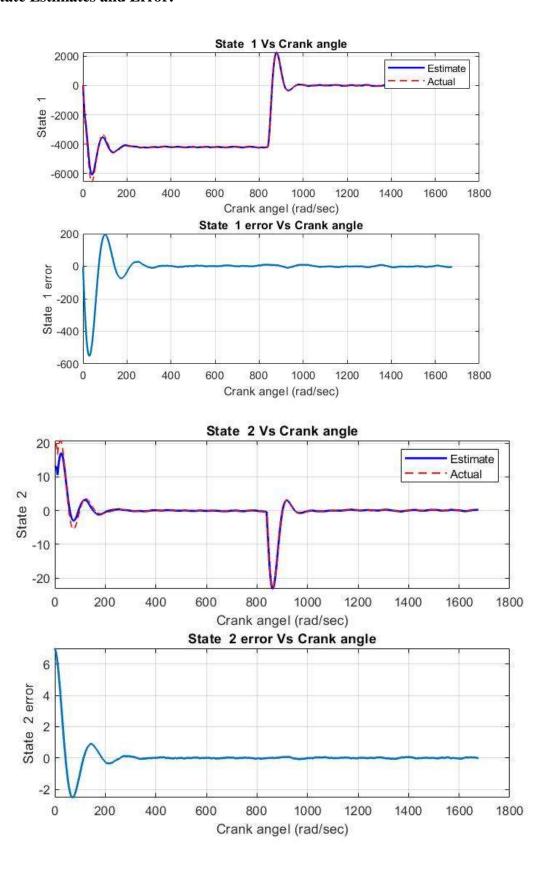
The overall response of the system with kalman filter is,

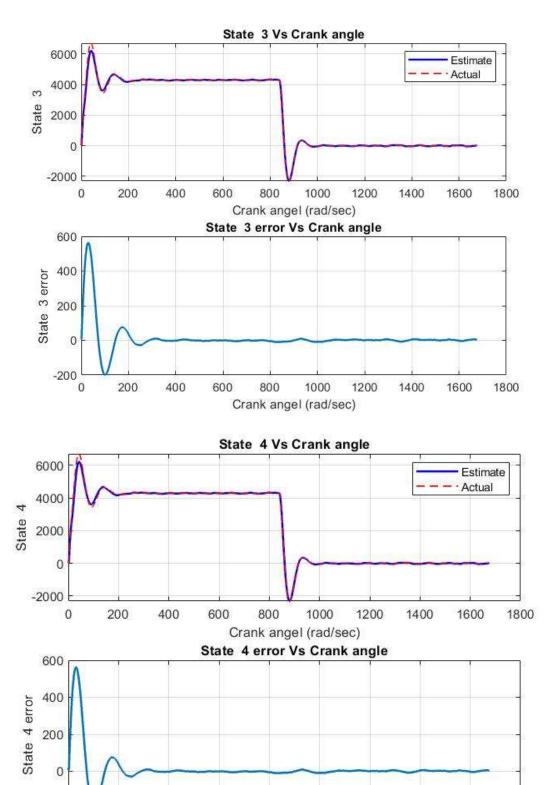


Closed Loop Response with Kalman Filter

NOTE: The speed in rpm starts at approx. 650rpm because the initial states is chosen non-zero for the plant also (to check regulatory response) and for the observer (to check convergence of the estimates) (both the initial conditions are not equal)

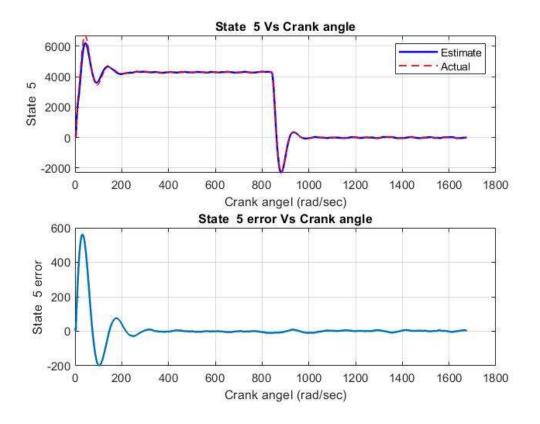
The State Estimates and Error:





Crank angel (rad/sec)

-200



Inference:

All the state were tracked significantly (response is satisfactory)

The throttle angle, overshoot, settling time are within desired region.

The steady state error is fairly zero (But chatters around zero value with small variation)