

# OBSERVER IMPLEMENTATION

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How exactly should we choose  $\bar{Q}$  and  $R$ ?

Define:

$$e = Cx + Du - y$$

$$Q = \text{diag}(q_1, q_2, \dots)$$

Suppose  $\sigma_i$  is the standard deviation of  $e_i$  (as you found in experiment). Then a good choice is:

$$q_i = (1/\sigma_i)^2 \quad \leftarrow \text{inverse of error variance}$$

Why? If  $\sigma_i$  is small, then  $q_i$  will be big, and so you are saying you trust the  $i$ th sensor a lot — which is exactly what you should do if your model of this sensor has low noise.

LQR is a standard way to choose  $L$

$$\text{minimize}_{d, n} \int_{-\infty}^0 (n(t)^T Q n(t) + d(t)^T R d(t)) dt$$

$$\text{subject to} \quad \begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t) + d(t) \\ y(t) &= Cx(t) + Du(t) + n(t) \end{aligned}$$

The solution is:

$$\hat{x} = A\hat{x} + Bu - L(C\hat{x} + Du - y)$$

where

$$L = \text{lqr}(A^T, C^T, \bar{Q}^{-1}, \bar{Q}^{-1})^T$$

$$K = \text{lqr}(A, B, Q, R)$$

$\uparrow \quad \uparrow$   
 $Q \text{ big} \Rightarrow n \text{ small} \Rightarrow \text{TRUST SENSORS}$   
 $R \text{ big} \Rightarrow d \text{ small} \Rightarrow \text{TRUST DYNAMIC MODEL}$

Choose  $R$  in exactly the same way, by looking at the error

$Ax + Bu - \dot{x}$   
 in the linearized EOMs.

$\leftarrow$  this is just intuition — it is possible to prove that this choice of  $\bar{Q}$  is, in fact, optimal

Is it possible to avoid converting between  $s, i, o$  and  $x, u, y$  when implementing an observer?

YES

This is an observer:

$$\dot{\hat{x}} = A\hat{x} + Bu - L(C\hat{x} - y)$$

$\frac{d}{dt}(\hat{s} - seq) = \dot{\hat{s}} - 0 = \dot{\hat{s}}$        $\hat{s} - seq$        $i - ieq$        $o - oeq$

This is an equivalent way to write the same observer:

$$\dot{\hat{s}} = A(\hat{s} - seq) + B(i - ieq) - L(C(\hat{s} - seq) + D(i - ieq) - (o - oeq))$$

↑  
this is often easier to implement in practice  
(it avoids the need to define  $x, u, y$  in code)

How do I update the estimate of a state that is not observable?

The best you can do is integrate:

$$\dot{\hat{x}} = A\hat{x} + Bu$$

For example, since  $o_x$  is not observable, then we must integrate the linearized EOM for  $o_x$ :

$$\dot{\hat{o}}_x = v_x$$

We approximate the integral as:

$$\hat{o}_x(t+\Delta t) = \hat{o}_x(t) + \Delta t v_x(t)$$

How do I update the estimate of a state that is observable?

We use our observer:

$$\dot{\hat{x}} = A\hat{x} + Bu - L(C\hat{x} + Du - y)$$

For example, consider  $o_2$  (an observable state). You should find the row of  $L$  corresponding to  $o_2$  is:

$$[0 \quad 0 \quad l]$$

for some number  $l$ . So, from the linearized EOM for  $o_2$  and the linearized measurement equation for  $r$ , we find:

$$\dot{\hat{o}}_2 = v_2 - l(o_2 - r)$$

We approximate the integral as:

$$\hat{o}_2(t+\Delta t) = \hat{o}_2(t) + \Delta t \left( v_2(t) - l \underbrace{(o_2(t) - r(t))} \right)$$

these error terms are often used more than once, so it is best to precompute them - we might call this one  $\tau_{er}$