

# Output Feedback

Adapted from Chapter 5 in *Adaptive Control (2nd Edition)* by K. J. Astrom and B. Wittenmark, Addison-Wesley, 1995

## 1. Adaptation of a Feedforward Gain

We now use Lyapunov theory to derive parameter adjustment laws for the problem of adjusting a feedforward gain. We consider the case in which the plant has transfer function  $k\hat{G}(s)$ , where  $\hat{G}(s)$  is known and  $k$  is unknown. The desired response is given by the transfer function  $k_0\hat{G}(s)$ . This problem was discussed previously in Lecture 4 (Example 1). The error is given by

$$e_0 = [k\hat{G}(s)\theta - k_0\hat{G}(s)]r = k\hat{G}(s)(\theta - \theta^*)r$$

where  $\theta^* = k_0/k$ . To use Lyapunov theory, we first introduce a state space representation of the transfer function  $\hat{G}$ . The relation between the parameter  $\theta$  and the error  $e_0$  can then be written as

$$\begin{aligned}\frac{dx}{dt} &= Ax + B(\theta - \theta^*)r \\ e_0 &= Cx.\end{aligned}\tag{1}$$

If the homogeneous system  $\dot{x} = Ax$  is asymptotically stable, there exist positive definite matrices  $P$  and  $Q$  such that

$$A^T P + PA = -Q.\tag{2}$$

Choose the following function as a candidate for a Lyapunov function:

$$v(x, \theta) = \frac{1}{2}gx^T Px + \frac{1}{2}(\theta - \theta^*)^2.$$

The time derivative of  $v$  along the differential equation, Eqs. (1), is given by

$$\frac{dv}{dt} = \frac{g}{2} \left( \frac{dx^T}{dt} Px + x^T P \frac{dx}{dt} \right) + (\theta - \theta^*) \frac{d\theta}{dt}.$$

Using Eqs. (1), we get

$$\begin{aligned}\frac{dv}{dt} &= \frac{g}{2} \left( (Ax + Br(\theta - \theta^*))^T Px + x^T P (Ax + Br(\theta - \theta^*)) \right) + (\theta - \theta^*) \frac{d\theta}{dt} \\ &= -\frac{g}{2} x^T Qx + (\theta - \theta^*) \left( \frac{d\theta}{dt} + grB^T Px \right).\end{aligned}$$

If the parameter adjustment law is chosen to be

$$\frac{d\theta}{dt} = -g r B^T P x \quad (3)$$

we find that the derivative of the Lyapunov function will be negative as long as  $x \neq 0$ . The state vector  $x$  and the error  $e_0 = Cx$  will go to zero as  $t$  goes to infinity. Notice, however, that the parameter error will not necessarily go to zero.

**Output feedback** The result obtained is quite restrictive because it requires that all state variables are known. A parameter adjustment law that uses output feedback can be obtained if the Lyapunov function can be chosen so that

$$B^T P = C$$

where  $C$  is the output matrix of the system in Eqs. (1). With this choice of  $P$  it follows that

$$B^T P x = Cx = e_0$$

and the adjustment rule becomes

$$\frac{d\theta}{dt} = -g r e_0.$$

The appropriate condition is given by the celebrated Kalman-Yakubovich lemma.

**Lemma 1 (Kalman-Yakubovich lemma)** (see also Lemma 2.6.2, p. 83, in the textbook by Sastry and Bodson): Let the time-invariant linear system

$$\begin{aligned} \frac{dx}{dt} &= Ax + Bu \\ y &= Cx \end{aligned}$$

be completely controllable and completely observable. The transfer function

$$\hat{G}(s) = C(sI - A)^{-1} B$$

is strictly positive real if and only if there exist positive definite matrices  $P$  and  $Q$  such that

$$A^T P + PA = -Q$$

and

$$B^T P = C.$$

**Theorem 1 (MRAS using the Lyapunov rule)**: Consider the problem of adapting a feedforward gain. Assume that the transfer function is strictly positive real. Then the parameter adjustment rule

$$\frac{d\theta}{dt} = -g r e_0 \quad (4)$$

where  $g$  is a positive constant, makes the output error  $e_0$  in Eqs. (1) go to zero.

The control law of Eq. (4) is very similar to the control law obtained by the MIT rule,  $d\theta/dt = -gy_m e_0$ . However, the adaptation gain  $g$  in the control law derived by the MIT rule needs to be small, while there is no such constraint on  $g$  in Eq. (4).

## 2. Output Feedback

We now present an outline for the derivation of an MRAS for adjusting the parameters of a control based on output feedback in a fairly general case. A process with one input and one output is considered. It is assumed that the dynamics are linear and that the control problem is formulated as model-following. The key assumption is that the controller can be parameterized in such a way that the error is linear in the controller parameters. The outline of the derivation of the MRAS is described as follows:

**Step 1:** Find a controller structure that admits perfect output tracking.

**Step 2:** Derive an error model of the form

$$\varepsilon = \hat{G}_1(s)\{w^T(t)(\theta - \theta^*)\} \quad (5)$$

where  $\hat{G}_1$  is a strictly positive real transfer function,  $\theta$  is the controller parameters, and  $\theta^*$  is the process parameters or the nominal value of  $\theta$ . The right-hand side should be expressed in computable quantities.

**Step 3:** Use the parameter adjustment law

$$\frac{d\theta}{dt} = -g w \varepsilon. \quad (6)$$

Notice that the error in Eq. (5) is linear in the parameters, a condition that imposes restrictions on the models and controllers that can be dealt with. A model of the form (5) is typically obtained by algebraic manipulations, filtering, and error augmentation.