

## \* Lyapunov - based MRAC Design:

- The main advantage of Lyapunov design is that it guarantees a stable closed-loop system.
- For a linear, asymptotically stable governed by a matrix  $\hat{A}$ , a positive symmetric matrix  $\hat{Q}$  yields a positive symmetric matrix 'P' by the equation:

$$\hat{A}^T P + P \hat{A} = -\hat{Q}$$

This equation is known as "Lyapunov's equation".

- The main drawback of Lyapunov design is that there is no symmetric way of finding a suitable Lyapunov function 'V' leading to a specific adaptive Law.
- For example, if one wants to add a proportional term to adaptive Law, it is not trivial to find the corresponding Lyapunov function.
- The hyperstability approach is more flexible than the Lyapunov approach.

## \* The Lyapunov Equation:

- Let the linear system :  $\dot{X} = A X$  be stable.
- Let ' $\hat{Q}$ ' be an arbitrary positive definite matrix.
- Then, the Lyapunov equation:  $\hat{A}^T P + P \hat{A} = -\hat{Q}$  always has a unique solution where 'P' is positive definite.
- The function  $V(X) = X^T P X$  is then a Lyapunov function.

### \* State Feedback:

- Process:  $\dot{X} = AX + Bu$
- Desired response:  $\dot{X}_m = A_m X_m + B_m u_c$
- Control Law:  $u = Mu_c - LX$
- closed-loop system:

$$\begin{aligned}\dot{X} &= AX + B [Mu_c - LX] \\ &= \underline{[A - BL]} X + \underline{BM} u_c \\ &= A_c(\theta) X + B_c(\theta) u_c\end{aligned}$$

- The parameterization is:  $A_c(\theta^0) = A_m$  and  $B_c(\theta^0) = B_m$
- For compatibility we need:  $A - A_m = BL$  and  $B_m = BM$

$$\text{or } L = (B^T B)^{-1} B^T [A - A_m]$$

$$M = (B^T B)^{-1} B^T B_m$$

### \* Error Equation:

The error  $e = X - X_m$  satisfies

$$\begin{aligned}\dot{e} &= \dot{X} - \dot{X}_m = AX + Bu - A_m X_m - B_m u_c \\ &= AX + B[Mu_c - LX] - A_m X_m - B_m u_c + \underline{A_m X - A_m X}\end{aligned}$$

Hence:

$$\begin{aligned}\dot{e} &= [A - A_m - BL] X + [BM - B_m] u_c + \underbrace{[X - X_m]}_e A_m \\ &= A_m e + [A - A_m - BL] X + [BM - B_m] u_c\end{aligned}$$

$$= A_m e + \underbrace{[A - BL - A_m]}_{A_c(\theta)} x + \underbrace{[B_M - B_m]}_{B_c(\theta)} u_c$$

$$= A_m e + [A_c(\theta) - \underbrace{A_m}_{A_c(\theta^0)}] x + [B_c(\theta) - \underbrace{B_m}_{B_c(\theta^0)}] u_c$$

$$= A_m e + \psi [\theta - \theta^0]$$

• Try:  $V(e, \theta) = \frac{1}{2} (\gamma e^T P e + [\theta - \theta^0]^T [\theta - \theta^0])$

• Differentiating V:

$$\dot{V} = \frac{\gamma}{2} [e^T P \dot{e} + \dot{e}^T P e + [\theta - \theta^0]^T \dot{\theta} + \overset{\substack{= \text{since} \\ \text{scalar}}}{\dot{\theta}^T [\theta - \theta^0]}]$$

$$= \frac{\gamma}{2} [e^T P (A_m e + \psi [\theta - \theta^0]) + e^T A_m^T P e + \psi^T [\theta - \theta^0] P e + [\theta - \theta^0]^T \dot{\theta}]$$

$$= \frac{\gamma}{2} [e^T P A_m e + e^T A_m^T P e + e^T P \psi [\theta - \theta^0] + \psi^T [\theta - \theta^0] P e + [\theta - \theta^0]^T \dot{\theta}]$$

• Now, consider the Lyapunov equation:  $A_m^T P + P A_m = -Q$

It is possible because reference model always stable.

$$\Rightarrow \dot{V} = -\frac{\gamma}{2} e^T Q e + \gamma [\theta - \theta^0]^T \psi^T P e + [\theta - \theta^0]^T \dot{\theta}$$

$$= -\frac{\gamma}{2} e^T Q e + [\theta - \theta^0]^T (\underbrace{\gamma \psi^T P e}_0 + \dot{\theta})$$

$$\Rightarrow \boxed{\dot{\theta} = -\gamma \psi^T P e}$$

Parameter adjustment

If parameter adjustment law is chosen as :

$$\dot{\hat{\theta}} = -\gamma \psi^T p e$$

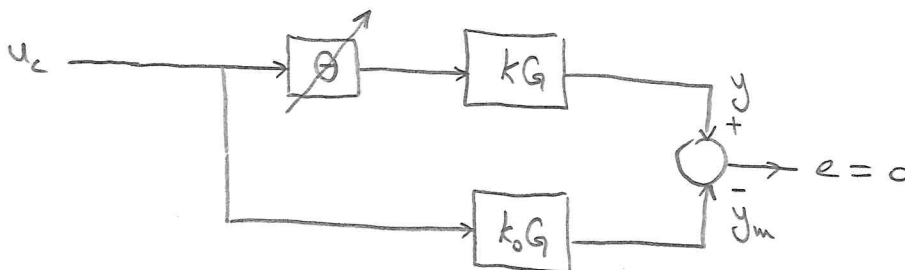
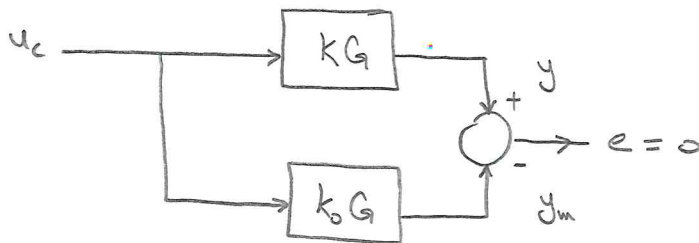
We obtain :

$$\dot{V} = -\frac{\gamma}{2} e^T Q e$$

which is negative sem-definite

- note that this assumes knowledge of the state.
- This does not mean that  $\hat{\theta} - \theta^*$  converges to zero, but  $e \rightarrow 0$ .

\* Adaptation of Feedforward Gain:



- Process :  $\hat{G} = k G(s)$
- Desired response :  $\hat{G}_w = k_0 G(s)$
- Control Law :  $u = \hat{\theta} u_c$

$$\begin{aligned}
 e &= y - y_m \\
 &= k G(p) u - k_o G(p) u_c \\
 &= k G(p) \theta u_c - k_o G(p) u_c \\
 &= k \left[ G(p) \theta u_c - \frac{k_o}{k} G(p) u_c \right] \quad , \quad \frac{k_o}{k} = \theta^o \\
 &= k G(p) u_c [\theta - \theta^o]
 \end{aligned}$$

with

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

$$\dot{X} = AX + \underbrace{\hat{B}k}_{B \theta} u_c \quad , \quad y = CX$$

$$y = \underbrace{C (sI - A)^{-1} B}_{G(s)} u \Rightarrow e = k C (sI - A)^{-1} \hat{B} u_c (\theta - \theta^o)$$

$$\Rightarrow \dot{X} = AX + B(\theta - \theta^o) u_c$$

$$e = CX \Rightarrow \dot{e} = C\dot{X}$$

A candidate Lyapunov function is :

$$V = \frac{1}{2} (\gamma X^T P X + (\theta - \theta^o)^2)$$

$$\dot{V} = \frac{\gamma}{2} (\underbrace{\dot{X}^T}_{\dot{X} = AX + B(\theta - \theta^o) u_c} P X + X^T P \dot{X}) + (\theta - \theta^o) \dot{\theta}$$

$$\begin{aligned}
 &= \frac{\gamma}{2} [X^T A^T P X + u_c (\theta - \theta^o) B^T P X + X^T P A X + X^T P B (\theta - \theta^o) u_c \\
 &\quad + (\theta - \theta^o) \dot{\theta}]
 \end{aligned}$$

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

Then,

$$\begin{aligned}\dot{V} &= -\frac{\gamma}{2} X^T Q X + \gamma (\theta - \theta^0) B^T P X u_c + (\theta - \theta^0) \dot{\theta} \\ &= -\frac{\gamma}{2} X^T Q X + (\theta - \theta^0) \underbrace{[\gamma B^T P X u_c + \dot{\theta}]}_0\end{aligned}$$

update Law  $\Rightarrow \boxed{\dot{\theta} = -\gamma u_c B^T P X}$

gives  $\dot{V} = -\frac{\gamma}{2} X^T Q X \quad X \rightarrow 0$

$e = C X$  as  $X \rightarrow 0$  then  $e \rightarrow 0$

if we can find  $P$  such that  $B^T P = C$

then, the adaptation law becomes

$$\boxed{\dot{\theta} = -\gamma u_c e}$$

i.e. this is now output feedback and the state  $X$  is not required.

### \* Definition:

- A rational transfer function 'G' with real coefficients is positive real (PR) if:  

$$\operatorname{Re}[G(s)] \geq 0 \quad \text{for} \quad \operatorname{Re}[s] \geq 0$$
- A rational transfer function 'G' with real coefficients is strictly positive real (SPR) if  $G(s-\epsilon)$  is PR for some  $\epsilon > 0$ .

$$\rightarrow G(s) = \frac{1}{s+1} \quad \text{is SPR.}$$

$$\rightarrow G(s) = \frac{1}{s} \quad \text{is PR but not SPR.}$$

Proof:

$$\begin{aligned} \bullet \quad G(s) = \frac{1}{s+1} &\Rightarrow G(j\omega) = \frac{1}{j\omega+1} \times \frac{1-j\omega}{1-j\omega} = \frac{1-j\omega}{\omega^2+1} \\ &= \underbrace{\frac{1}{\omega^2+1}}_{\text{Real}} - j \underbrace{\frac{\omega}{\omega^2+1}}_{\text{Imag.}} \end{aligned}$$

$$\Rightarrow \operatorname{Re}[G(j\omega)] = \frac{1}{\omega^2+1} > 0 \quad \forall \omega, \text{ then SPR.}$$

$$\bullet \quad G(s) = \frac{1}{s} \Rightarrow G(j\omega) = \frac{1}{j\omega} \times \frac{-j\omega}{-j\omega} = -j \frac{\omega}{\omega^2} = -j \frac{1}{\omega}, \quad \text{no real part}$$

$$\Rightarrow \operatorname{Re}[G(j\omega)] \geq 0 \quad \forall \omega, \text{ then PR.}$$

Theorem [Kalman - Yakubovich Lemma] :

Let the Linear time-invariant system :

$$\dot{X} = AX + Bu$$

$$y = Cx$$

be completely controllable and completely observable. The transfer function

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

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

$$\begin{aligned} A^T P + PA &= -Q \\ B^T P &= C \end{aligned}$$