

EML 6351

## Non linear Control - 2 - Adaptive Control.

### Project - 1.

#### Dynamics.

$$M(\phi)\ddot{\phi} + C(\phi, \dot{\phi}) + G(\phi) = \tau$$

$$\text{where } \phi = \begin{bmatrix} \phi_1 \\ \phi_2 \end{bmatrix} \quad \dot{\phi} = \begin{bmatrix} \dot{\phi}_1 \\ \dot{\phi}_2 \end{bmatrix} \quad \ddot{\phi} = \begin{bmatrix} \ddot{\phi}_1 \\ \ddot{\phi}_2 \end{bmatrix} \quad \tau = \begin{bmatrix} \tau_1 \\ \tau_2 \end{bmatrix}$$

$$M(\phi) = \begin{bmatrix} (m_1 l_1^2 + m_2(l_1^2 + 2l_1 l_2 c_2 + l_2^2)) & m_2(l_1 l_2 c_2 + l_2^2) \\ m_2(l_1 l_2 c_2 + l_2^2) & m_2 l_2^2 \end{bmatrix}$$

$$C(\phi, \dot{\phi}) = \begin{bmatrix} -2m_2 l_1 l_2 s_2 \dot{\phi}_1 \dot{\phi}_2 \\ 0 \end{bmatrix} + \begin{bmatrix} -m_2 l_1 l_2 s_2 \dot{\phi}_2^2 \\ m_2 l_1 l_2 s_2 \dot{\phi}_1^2 \end{bmatrix}$$

$$G(\phi) = \begin{bmatrix} (m_1 + m_2) g l_1 c_1 + m_2 g l_2 c_2 \\ m_2 g l_2 c_2 \end{bmatrix}$$

Defining tracking error.

$$e = \phi_d - \phi$$

$$\dot{e} = \dot{\phi}_d - \dot{\phi}$$

$$\ddot{e} = \ddot{\phi}_d - \ddot{\phi}$$



$$M(\phi) \ddot{e} = M(\phi) \ddot{\phi}_0 - M(\phi) \ddot{\phi} \quad \text{--- (1)}$$

filtered tracking error

$$\tau = \dot{e} + \alpha e$$

$$\dot{\tau} = \ddot{e} + \alpha \dot{e}$$

$$\Rightarrow M(\phi) \dot{\tau} = M(\phi) \ddot{e} + M(\phi) \alpha \dot{e}$$

Sub eqn (1) we get

$$M(\phi) \dot{\tau} = (M(\phi) \ddot{\phi}_0 - M(\phi) \ddot{\phi}) + M(\phi) \alpha \dot{e}$$

From dynamics

$$M(\phi) \ddot{\phi} = \tau - c(\phi, \dot{\phi}) - G(\phi)$$

$$\begin{aligned} \Rightarrow M(\phi) \dot{\tau} &= M(\phi) \ddot{\phi}_0 - (\tau - c(\phi, \dot{\phi}) - G(\phi)) + M(\phi) \alpha \dot{e} \\ &= M(\phi) (\ddot{\phi}_0 + \alpha \dot{e}) + c(\phi, \dot{\phi}) + G(\phi) - \tau \end{aligned}$$

Define  $\theta \in \mathbb{R}^p$  where  $p = \text{no of terms in } \theta$

$$\text{Let } \tilde{\theta} = \theta - \hat{\theta}$$

$$\Rightarrow \tilde{\theta}^* = -\hat{\theta}^* \quad (\because \theta \text{ is a constant})$$

$\hat{\theta}(t) \in \mathbb{R}^p$  is the parametric update law.



Let  $\xi_1 = \begin{bmatrix} e \\ \dot{r} \\ \tilde{\theta} \end{bmatrix}$  be the stacked errors

Lyapunov candidate.

$$V(\xi_1, t) = \frac{1}{2} \tilde{e}^T \tilde{e} + \frac{1}{2} \dot{r}^T M(\phi) \dot{r} + \frac{1}{2} \tilde{\theta}^T \Gamma^{-1} \tilde{\theta}.$$

$$\dot{V} = \tilde{e}^T \dot{\tilde{e}} + \frac{1}{2} \dot{r}^T \dot{M}(\phi) \dot{r} + \frac{1}{2} \dot{r}^T M(\phi) \ddot{r} + \tilde{\theta}^T \Gamma^{-1} \dot{\tilde{\theta}}.$$

we know  $\dot{\tilde{\theta}} = -\dot{\hat{\theta}}$  and  $\dot{\tilde{e}} = \dot{r} - \alpha \tilde{e}$ .

$$\dot{V} = \tilde{e}^T (\dot{r} - \alpha \tilde{e}) + \frac{1}{2} \dot{r}^T \dot{M}(\phi) \dot{r} + \frac{1}{2} \dot{r}^T M(\phi) \ddot{r} - \tilde{\theta}^T \Gamma^{-1} \dot{\hat{\theta}}.$$

$$\dot{M}(\phi) = \frac{\partial}{\partial t} M(\phi).$$

$$= \frac{\partial}{\partial t} \begin{bmatrix} (m_1 l_1^2 + m_2 (l_1^2 + 2 l_1 l_2 c_2 + l_2^2)) & m_2 (l_1 l_2 (c_2 + l_2^2)) \\ m_2 (l_1 l_2 (c_2 + l_2^2)) & m_2 l_2^2 \end{bmatrix}$$

$$= \begin{bmatrix} -m_2 (2 l_1 l_2 s_2) \dot{\phi}_2 & -m_2 l_1 l_2 \dot{s}_2 \dot{\phi}_2 \\ -m_2 l_1 l_2 s_2 \dot{\phi}_2 & 0 \end{bmatrix}$$

$$\text{and } M(\phi) \ddot{r} = M(\phi) (\ddot{\phi}_0 + \alpha \dot{\tilde{e}}) + C(\phi, \dot{\phi}) + G(\phi) - \tau.$$

$$\dot{V} = \tilde{e}^T (\dot{r} - \alpha \tilde{e}) + \frac{1}{2} \dot{r}^T \dot{M}(\phi) \dot{r} + \frac{1}{2} \dot{r}^T M(\phi) \ddot{r} -$$

$$+ \frac{1}{2} \dot{r}^T (M(\phi) (\ddot{\phi}_0 + \alpha \dot{\tilde{e}}) + C(\phi, \dot{\phi}) + G(\phi) - \tau) - \tilde{\theta}^T \Gamma^{-1} \dot{\hat{\theta}}.$$



$$v = -\theta^T \alpha e + e^T r - \theta^T \Gamma^T \hat{\theta} \\ + r^T (M(\phi) (\tilde{\phi}_0 + \alpha e) + C(\phi, \phi) + G(\phi) + \frac{1}{2} M(\phi) r - \tau)$$

To find the regressor we want.

$$M(\phi) (\tilde{\phi}_0 + \alpha e) + C(\phi, \phi) + G(\phi) + \frac{1}{2} M(\phi) r = Y\theta$$

to get the minimum realization of  $\theta$ .

$$M(\phi) \nu = \begin{bmatrix} \underbrace{m_1 l_1^2}_{\theta_1} + \underbrace{m_2 l_1^2 + m_2 l_2^2 + 2 l_1 l_2 c_2}_{\theta_2} & \underbrace{m_2 l_1 l_2 c_2}_{\theta_2} + \underbrace{m_2 l_2^2}_{\theta_3} \\ \underbrace{m_2 l_1 l_2 c_2}_{\theta_2} + \underbrace{m_2 l_2^2}_{\theta_3} & \underbrace{m_2 l_2^2}_{\theta_3} \end{bmatrix} \nu$$

$$M(\phi) \nu = \begin{bmatrix} \theta_1 + 2\theta_2 c_2 & c_2 \theta_2 + \theta_3 \\ c_2 \theta_2 + \theta_3 & \theta_3 \end{bmatrix} \begin{bmatrix} \nu_1 \\ \nu_2 \end{bmatrix}$$

$$M(\phi) \nu = \begin{bmatrix} (\theta_1 + \theta_2) \nu_1 + (\theta_2 + \theta_3) \nu_2 \\ (\theta_2 + \theta_3) \nu_1 + \theta_3 \nu_2 \end{bmatrix}$$

$$-C(\phi, \phi) = f$$

$$M(\phi) \nu = \begin{bmatrix} \nu_1 & \nu_1 + \nu_2 \end{bmatrix}$$



$$M(\phi)N = \begin{bmatrix} (\theta_1 + 2\theta_2 c_2)N_1 + (\theta_2 c_2 + \theta_3)N_2 \\ (c_2 \theta_2 + \theta_3)N_1 + \theta_3 N_2 \end{bmatrix}$$

$$M(\phi)N = \begin{bmatrix} N_1 & 2N_1 c_2 + N_2 \theta c_2 & N_2 \\ 0 & c_2 N_1 & N_1 + N_2 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \end{bmatrix}$$

Now, looking at  $C(\phi, \phi^\circ)$ .

$$C(\phi, \phi^\circ) = \begin{bmatrix} -2m_2 l_1 l_2 s_2 \phi_1^\circ \phi_2^\circ - m_2 l_1 l_2 s_2 \phi_2^{\circ 2} \\ m_2 l_1 l_2 s_2 \phi_1^{\circ 2} \end{bmatrix}$$

$$= \begin{bmatrix} 0 & -2s_2 \phi_1^\circ \phi_2^\circ - s_2 \phi_2^{\circ 2} & 0 \\ 0 & s_2 \phi_1^{\circ 2} & 0 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \end{bmatrix}$$

Now looking at  $G(\phi)$ .

$$G(\phi) = \begin{bmatrix} (m_1 + m_2)g l_1 c_1 + m_2 g l_2 c_{12} \\ m_2 g l_2 c_{12} \end{bmatrix}$$

$$= \begin{bmatrix} g c_1 & g c_{12} \\ 0 & g c_{12} \end{bmatrix} \begin{bmatrix} \theta_4 \\ \theta_5 \end{bmatrix}$$



Now  $\dot{M}(\phi, \dot{\phi})\gamma + M(\phi, \dot{\phi})\ddot{\phi} = M(\phi, \dot{\phi})\ddot{\phi}$

$$\dot{M}(\phi, \dot{\phi})\gamma = \begin{bmatrix} -2m_2 l_1 l_2 s_2 \dot{\phi}_2^2 & -m_2 l_1 l_2 s_2 \dot{\phi}_2^2 \\ -m_2 l_1 l_2 s_2 \dot{\phi}_2^2 & 0 \end{bmatrix} \begin{bmatrix} \gamma_1 \\ \gamma_2 \end{bmatrix}$$

$$= \begin{bmatrix} -2\theta_2 \dot{\phi}_2^2 \gamma_1 & -\theta_2 s_2 \dot{\phi}_2^2 \gamma_2 \\ -\theta_2 s_2 \dot{\phi}_2^2 \gamma_1 & 0 \end{bmatrix}$$

$$\dot{M}(\phi, \dot{\phi})\gamma = \begin{bmatrix} 0 & -2\dot{\phi}_2^2 \gamma_1 - s_2 \dot{\phi}_2^2 \gamma_2 & 0 \\ 0 & -s_2 \dot{\phi}_2^2 \gamma_1 & 0 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \end{bmatrix}$$

Writing all the above

$$M(\phi)\ddot{\phi} = \begin{bmatrix} N_1 & 2C_1 N_1 + C_2 N_2 & N_2 & 0 & 0 \\ 0 & C_2 N_1 & N_1 + N_2 & 0 & 0 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \\ \theta_4 \\ \theta_5 \end{bmatrix}$$

$$C(\phi, \dot{\phi})\dot{\phi} = \begin{bmatrix} 0 & -2s_2 \dot{\phi}_1 \dot{\phi}_2^2 - s_2 \dot{\phi}_2^2 & 0 & 0 & 0 \\ 0 & s_2 \dot{\phi}_1^2 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \\ \theta_4 \\ \theta_5 \end{bmatrix}$$



$$G(\phi) = \begin{bmatrix} 0 & 0 & 0 & g_{C1} & g_{C12} \\ 0 & 0 & 0 & 0 & g_{C12} \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \\ \theta_4 \\ \theta_5 \end{bmatrix}$$

$$m(\phi, \phi^*) r = \begin{bmatrix} 0 & -2s_1 \phi_2^* r_1 - s_2 \phi_2^* r_2 & 0 & 0 & 0 \\ 0 & -s_2 \phi_2^* r_1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \\ \theta_4 \\ \theta_5 \end{bmatrix}$$

$$Y\theta = (Y_m(\phi, n) + Y_c(\phi, \phi^*) + Y_G(\phi) + Y_m(\phi, \phi^*, r)) \theta.$$

Now,

$$v(\xi, t) = -e^T \alpha e + e^T r + r^T (Y\theta - \tau) - \tilde{\theta}^T \Gamma^T \hat{\theta}^*$$

$$\text{Design, } \tau = Y\hat{\theta} + e + \beta r.$$

$$\dot{v} = -e^T \alpha e + r^T (Y\theta + e - (Y\hat{\theta} + e + \beta r)) - \tilde{\theta}^T \Gamma^T \hat{\theta}^*$$

$$\dot{v} = -e^T \alpha e - r^T \beta r + r^T Y\tilde{\theta} - \tilde{\theta}^T \Gamma^T \hat{\theta}^*$$

$$\text{design } \hat{\theta}^* = \Gamma^T Y^T r.$$

$$\dot{v} = -e^T \alpha e - r^T \beta r + \tilde{\theta}^T Y^T r - \tilde{\theta}^T \Gamma^T (\Gamma Y^T r)$$

$$\dot{v} = -e^T \alpha e - r^T \beta r.$$

$$\Rightarrow \dot{v} \leq -\underline{\alpha} \|e\|^2 - \underline{\beta} \|r\|^2.$$

where  $\underline{\alpha}$  and  $\underline{\beta}$  are min eigen values of  $\alpha$  and  $\beta$ .



Using Barbalat's Lemma, we can show that

$$\text{as } t \rightarrow \infty \quad e \rightarrow 0$$

$$\Rightarrow r = e + \dot{e} \Rightarrow r \rightarrow 0$$

$$\Rightarrow \phi_0 + \dot{e} \in L_2 \text{ as } \phi_0 \text{ is a constant.}$$

$$\Rightarrow \dot{N} \in L_2.$$

$$\Rightarrow \phi \in L_2 \text{ as } \phi_0, e \in L_2, \dot{\phi}, \ddot{\phi} \in L_2.$$

$$\Rightarrow Y_m(\phi), Y_c(\phi, \dot{\phi}), Y_g(\phi), Y_n(\phi, \dot{\phi}) \in L_2.$$

$$\Rightarrow Y \in L_2.$$

$$\hat{\theta} = Y^+ Y \text{ implies } \hat{\theta} \in L_2.$$

$$\tau = Y\hat{\theta} + \beta r + e \Rightarrow \tau \in L_2.$$

$$\dot{r} \in L_2 \text{ and } r \text{ is uniformly continuous.}$$

$$\Rightarrow \lim_{t \rightarrow \infty} r(t) = 0.$$

$e$  and  $r$  have Global asymptotic tracking.

Since  $r, e \in L_2$  & are Uniformly Continuous.

Therefore, the system has GAT-using Barbalat's Lemma.