

Content

- Adaptive Control of Manipulators (not included in exam)

Adaptive Control

- In the previous section, we saw that if the model parameters are correct, then we achieve zero tracking error:

$$\ddot{E} + K_v \dot{E} + K_p E = 0$$

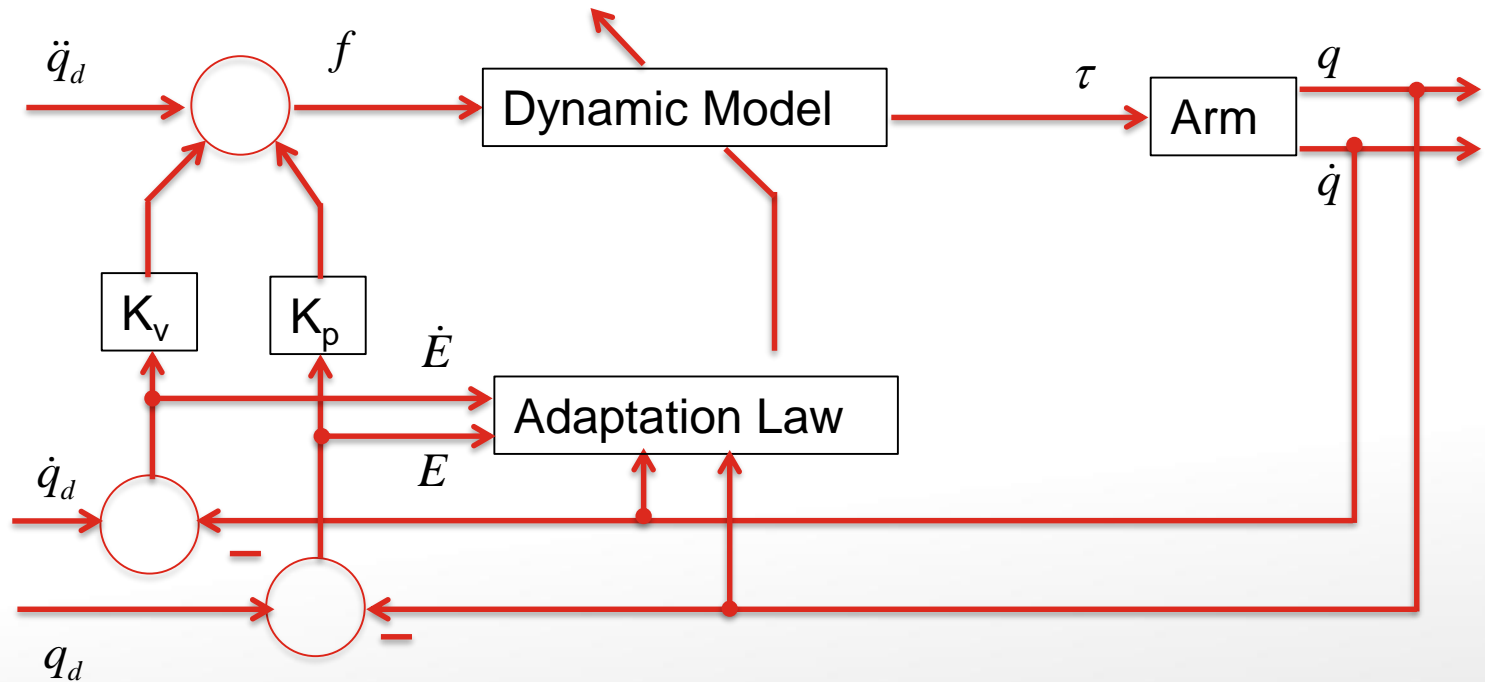
- However, if the model parameters do not match the real parameters, this will result in servo errors:

$$\ddot{E} + K_v \dot{E} + K_p E = \hat{M}^{-1} \left[(M - \hat{M}) \ddot{q} + (V - \hat{V}) + (G - \hat{G}) + (F - \hat{F}) \right]$$

- The adaptive control is based on the idea that the **model parameters could be continuously updated** until the servo error diminishes.

Adaptive Control

- One adaptive control scheme is as follows:



- If there is servo error (E and \dot{E}), the adaptation law will adjust the parameters in the dynamic model, until the error disappears.
- The system learns its own dynamic properties.

Indirect Adaptive Control

- Let's derive **one adaptive control scheme** here. (There are many other algorithms and this is just one of them).
- The manipulator's dynamic equation is:

$$M(q)\ddot{q} + V(q, \dot{q}) + G(q) = \tau$$

- The **model structure** is known.
- However, the **parameters** are not accurately known (or they may change).
- Now, the land hand side of the equation can be written in a **Linear-in-the-Parameters** form:

$$M(q)\ddot{q} + V(q, \dot{q}) + G(q) = \phi^T \theta = \tau$$

- where θ is the unknown **parameters**, and
- ϕ is the known **regressor**.

Indirect Adaptive Control

- Aside: What does **linear-in-parameters** mean?
- It means the parameters can be separated into a vector.
- E.g. 2R robot:

$$\begin{aligned}\tau_1 &= m_2 L_2^2 (\ddot{\theta}_1 + \ddot{\theta}_2) + m_2 L_1 L_2 c_2 (2\ddot{\theta}_1 + \ddot{\theta}_2) + (m_1 + m_2) L_1^2 \ddot{\theta}_1 - m_2 L_1 L_2 s_2 \dot{\theta}_2^2 \\ &\quad - 2m_2 L_1 L_2 s_2 \dot{\theta}_1 \dot{\theta}_2 + (m_1 + m_2) g L_1 c_1 + m_2 g L_2 c_{12} \\ \tau_2 &= m_2 L_1 L_2 c_2 \ddot{\theta}_1 + m_2 L_1 L_2 s_2 \dot{\theta}_1^2 + m_2 g L_2 c_{12} + m_2 L_2^2 (\ddot{\theta}_1 + \ddot{\theta}_2)\end{aligned}$$

$$\begin{bmatrix} \tau_1 \\ \tau_2 \end{bmatrix} = \underbrace{\begin{bmatrix} (\ddot{\theta}_1 + \ddot{\theta}_2) & c_2(2\ddot{\theta}_1 + \ddot{\theta}_2) - s_2 \dot{\theta}_2^2 - 2s_2 \dot{\theta}_1 \dot{\theta}_2 & \ddot{\theta}_1 & g c_1 & g c_{12} \\ (\ddot{\theta}_1 + \ddot{\theta}_2) & c_2 \ddot{\theta}_1 + s_2 \dot{\theta}_1^2 & 0 & 0 & g c_{12} \end{bmatrix}}_{\varphi^T} \underbrace{\begin{bmatrix} m_2 L_2^2 \\ m_2 L_1 L_2 \\ (m_1 + m_2) L_1^2 \\ (m_1 + m_2) L_1 \\ m_2 L_2 \end{bmatrix}}_{\theta}$$

Indirect Adaptive Control

- The **initially-inaccurate model** can be written as:

$$\hat{M}(q)\ddot{q} + \hat{V}(q, \dot{q}) + \hat{G}(q) = \phi^T \hat{\theta}$$

- where the “hat” symbol means estimates.
- We define:
 - $\tilde{\theta} = \theta - \hat{\theta}$ as the **parameter estimation error**.
 - $e = \phi^T \theta - \phi^T \hat{\theta} = \phi^T \tilde{\theta}$ as the **model error**.
 - This model error is available to us, because:
 - $M(q)\ddot{q} + V(q, \dot{q}) + G(q) = \phi^T \theta = \tau$ is the torque values at the joints, which can be measured.
 - $\hat{M}(q)\ddot{q} + \hat{V}(q, \dot{q}) + \hat{G}(q) = \phi^T \hat{\theta}$ is calculated by multiplying the measured regressor (joint acceleration, velocity, angles) with the parameter estimates.

Indirect Adaptive Control

- How can we update the parameter estimates $\hat{\theta}$ such that it gets closer and closer to θ ?
- We will use the so-called **Lyapunov method** as a design tool.
- What is it?

Lyapunov Stability Analysis

- In the 19th century, a Russian mathematician, **Lyapunov**, introduced the method of **deducing the stability** of a system, by analyzing its **energy** and the **rate of change** of the energy.
- The idea is intuitive:
 - Assume the **energy of a system is always non-negative (positive or zero)**.
 - If the **rate of change of energy is negative**, it means the system is losing energy.
 - System will slow down and eventually stops.
 - **Asymptotically stable!**
 - If the **rate of change of energy is positive**, then the is gaining energy.
 - System will speed up and eventually grow unbounded.
 - **Unstable!**

Lyapunov Stability Analysis

- Example: mass-spring-damper system: $m\ddot{x} + b\dot{x} + kx = 0$
- The total energy of the system, V , is given by:

$$V = \frac{1}{2}m\dot{x}^2 + \frac{1}{2}kx^2$$

- Note that this **energy is always non-negative**.
- The rate of change of the energy is obtained by differentiating V wrt. time:

$$\dot{V} = m\dot{x}\ddot{x} + kx\dot{x} = \dot{x}(m\ddot{x} + kx)$$

- Substituting the first equation into \dot{V} gives:


$$\dot{V} = \dot{x}(m\ddot{x} + kx) = \dot{x}(-b\dot{x}) = -b\dot{x}^2$$

- The **rate of change of energy is always negative** (for positive b), except that it becomes zero when \dot{x} becomes zero.
- This implies that the **system will lose energy and come to a stop**.

Indirect Adaptive Control

- Coming back to our system:
- Let:
$$V = \frac{1}{2} \tilde{\theta}^T \Gamma^{-1} \tilde{\theta}$$
- where $\Gamma = \Gamma^T > 0$ is a positive definite matrix, which determines the **learning rate** of the parameters.
 - Larger entries would make learning faster, but the parameter updates may become oscillatory.
 - Need some trade-offs.

Indirect Adaptive Control

- The derivatives of V is: $\dot{V} = \tilde{\theta}^T \Gamma^{-1} \dot{\tilde{\theta}}$
- Our aim is to design the updates of $\hat{\theta}$ such that \dot{V} becomes non-positive.
- If we design: $\dot{\tilde{\theta}} = -\dot{\hat{\theta}} = -\Gamma \phi e$


When model error is big, the parameters will be updated fast.
- Then:

$$\begin{aligned} \dot{V} &= \tilde{\theta}^T \Gamma^{-1} (-\Gamma \phi e) \\ &= -\underbrace{\tilde{\theta}^T \phi}_{e^T} e \\ &= -2e^T e \end{aligned}$$
- V -dot is non-positive, thus: $\frac{1}{2} \tilde{\theta}(t)^T \Gamma^{-1} \tilde{\theta}(t) \leq \frac{1}{2} \tilde{\theta}(0)^T \Gamma^{-1} \tilde{\theta}(0)$
 - The estimates are likely to become better!

Indirect Adaptive Control

- Finally, as the parameter estimates are being updated, we use the values into any control techniques, as if the parameters are correct.
 - This is called the **Certainty Equivalence Principle**.

- Summary:

- Manipulator dynamics: $M(q)\ddot{q} + V(q, \dot{q}) + G(q) = \varphi^T \theta = \tau$

- Model with estimated parameters: $\hat{M}(q)\ddot{q} + \hat{V}(q, \dot{q}) + \hat{G}(q) = \varphi^T \hat{\theta}$

- Update the parameters as: $\dot{\hat{\theta}} = \Gamma \varphi e$

- where $e = \varphi^T \theta - \varphi^T \hat{\theta} = \varphi^T \tilde{\theta}$

- Use the estimated parameters in the control law, for e.g.

$$\tau = \hat{M}(q)f + \hat{V}(q, \dot{q}) + \hat{G}(q) \quad f = \ddot{q}_d + K_v \dot{E} + K_p E$$

Thank you!

Have a good evening.

