

29.10.25

ПРОБЛЕМЫ ФИЗИЧЕСКОГО МИРА:

- АДСКР. ВРЕМЕНУ (+ измерений)

- ЗАДЕРЖКИ

- ШУМЫ

- ВОЗМУЩЕНИЯ

- НЕТОЧНОСТИ МОДЕЛИ $\dot{x} = f(x, u)$; $\dot{x} = f^M(x, u)$

АДАПТИВНОСТЬ

$$\dot{x} = f^M(x, u)$$

\hat{M} - ???

$$\|f^M(x, u) - \hat{f}^M(x, u)\| \xrightarrow{\hat{M}} \min$$

РОБАСТНОСТЬ

$$\dot{x} = f^M(x, u)$$

$$u = u(x(t)) \text{ m.u. } \int_0^T \dots dt \rightarrow \min \text{ (ОБЫЧН.)}$$

$$\max_M \int_0^T \dots dt \xrightarrow{u} \min$$

МИНИМ. СТОИМ. В ХУДШ. СЛУЧАЕ

α CLF (adaptive CLF)

$$\ddot{\theta} = \frac{g}{l} \sin \theta + \frac{\tau}{ml^2} - M\dot{\theta}$$

КАК ПРОИЗВ. КОМПЕНСАЦИЮ?

$$\tau = -\tau_{\max} \operatorname{sign}((E - E_d)\dot{\theta}) \quad (\text{БЫЛО})$$

$$\tau_b = -0,8 \tau_{\max} \operatorname{sign}((E - E_d)\dot{\theta}) \quad (\text{СТАЛО})$$

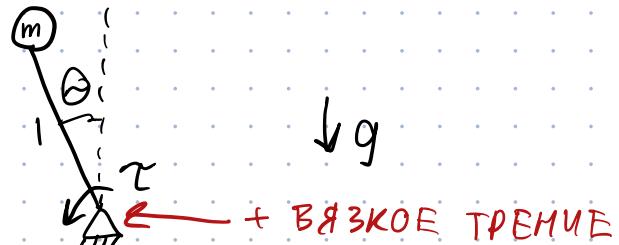
τ_b baseline τ_c compensatory

$$\tau = \tau_b + \tau_c$$

Если M известно ($\hat{M} = M$), то

$$\frac{\tau}{ml^2} - M\dot{\theta} = \frac{\tau_b}{ml^2}$$

$$\tau_b + \tau_c - ml^2 M\dot{\theta} = \tau_b$$



$$x = \begin{pmatrix} \theta \\ \dot{\theta} \end{pmatrix}; u = (\tau)$$

$$\tau_c = ml^2 \mu \dot{\theta}$$

ЕСЛИ $\hat{m} \neq M$, ТО

$$\tau_c = ml^2 \hat{\mu} \dot{\theta}$$

КАКИМ СТАНЕТ УР. ДИНАМИКИ ПРИ НЕИДЕАЛЬНОЙ КОМПЕНС.?

$$\tau = \tau_b + ml^2 \hat{\mu} \dot{\theta}$$

$$\ddot{\theta} = \frac{g}{l} \sin \theta + \frac{\tau_b}{ml^2} + \frac{ml^2 \hat{\mu} \dot{\theta}}{ml^2} - \mu \dot{\theta} = \frac{g}{l} \sin \theta + \frac{\tau_b}{ml^2} + (\hat{\mu} - \mu) \dot{\theta}$$

ВВЕДЕМ АДАПТИВН. Ф-Ю УПР. ЛЯПУНОВА:

$$V = \frac{(E - E_0)^2}{2} + \frac{(\hat{\mu} - \mu)^2}{2\lambda}$$

$$E = \frac{ml^2 \dot{\theta}^2}{2} + mg/l \cos \theta$$

// ЕД. ИЗМ. КОЭФФ. ВЯЗ. ТР.

ДЛЯ МАЧТН.:

H·M·C
PAA

$$\frac{dV}{dt} = (E - E_0)(ml^2 \dot{\theta} \ddot{\theta} - mg \sin \theta \dot{\theta}) + \frac{(\hat{\mu} - \mu)}{\lambda} \dot{\mu} =$$

$$= (E - E_0) \dot{\theta} \left(ml^2 \frac{g}{l} \sin \theta + ml^2 \cdot \frac{\tau_b}{ml^2} + ml^2 (\hat{\mu} - \mu) \dot{\theta} - mg \sin \theta \right) +$$

$$+ \frac{(\hat{\mu} - \mu)}{\lambda} \dot{\mu} = (E - E_0) \dot{\theta} \tau_b + (E - E_0) \dot{\theta} ml^2 (\hat{\mu} - \mu) \dot{\theta} + \frac{(\hat{\mu} - \mu)}{\lambda} \dot{\mu} =$$

$$= \underbrace{(E - E_0) \dot{\theta} \tau_b}_{< 0} + \underbrace{\frac{(\hat{\mu} - \mu)}{\lambda} ((E - E_0) \dot{\theta}^2 ml^2 + \frac{\dot{\mu}}{\lambda})}_{\neq 0} < 0$$



$$(E - E_0) \dot{\theta}^2 ml^2 + \frac{\dot{\mu}}{\lambda} = 0$$

$$\dot{\mu} = -\lambda (E - E_0) \dot{\theta}^2 ml^2$$

$$\mu_{\text{hat}} - \mu_{\text{hat_dot}} * dt$$

ИДЕНТИФИКАЦИЯ ЛИНЕЙНОЙ СИСТЕМЫ

$$x_{k+1} = Ax_k$$

$$\{x_i\}_{i=0}^N$$

$$x_1 = Ax_n$$

1)

$$\begin{array}{|c|c|c|c|c|} \hline x_1 & x_2 & \dots & \dots & x_N \\ \hline \end{array} = \begin{array}{|c|} \hline A \\ \hline \end{array} \begin{array}{|c|c|c|} \hline x_0 & x_1 & \dots \\ \hline \end{array}$$

x_1 x_0

$$X_1 = Ax_0 \quad | \cdot X_0^\top$$

$$X_1 X_0^\top = A X_0 X_0^\top \quad | \cdot (X_0 X_0^\top)^{-1}$$

$$A = X_1 X_0^\top (X_0 X_0^\top)^{-1}$$

$$x_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$x_2 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

$$X_0 = \begin{pmatrix} 0 & 1 & 2 & 3 & 4 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix}$$

$$X_1 = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix}$$

$$A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 & 2 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix} = \begin{pmatrix} 5 & 3 \\ 3 & 3 \end{pmatrix}$$

2) $x_{k+1} = Ax_k$ \hat{A}

$$\hat{x}_{k+1} = \hat{A} x_k$$

$$e_{k+1} = x_{k+1} - \hat{x}_{k+1} = (A - \hat{A}) x_k = x_{k+1} - \hat{A} x_k$$

$$\frac{\partial e_{k+1}^T e_{k+1}}{\partial \hat{A}} = \frac{\partial (x_{k+1} - \hat{A} x_k)^T (x_{k+1} - \hat{A} x_k)}{\partial \hat{A}} =$$

$$= \frac{\partial}{\partial \hat{A}} \left(x_{k+1}^\top x_{k+1} - 2x_{k+1}^\top \hat{A} x_k + x_k^\top \hat{A}^\top \hat{A} x_k \right) =$$

$$= -2x_k^\top x_{k+1}^\top + 2 \hat{A} x_k x_k^\top$$

$$\hat{A}^- = \cancel{A} \frac{\partial e^\top e}{\partial \hat{A}}$$

$$x_{k+1} = A x_k \rightsquigarrow x_{k+1} = A x_k + B u_k$$