

29.10.25

ПРОБЛЕМЫ ФИЗИЧЕСКОГО МИРА:

- ДИСКР. ВРЕМЕНИ (+ ИЗМЕРЕНИЙ)
- ЗАДЕРЖКИ
- ШУМЫ
- ВОЗМУЩЕНИЯ
- НЕТОЧНОСТИ МОДЕЛИ

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

АДАПТИВНОСТЬ

VS

РОБАСТНОСТЬ

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

$\hat{M} - ???$

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

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

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

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

МИНИМ. СТОИМ. В ХУДШ. СЛУЧАЕ

α CLF (adaptive CLF)

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

КАК ПРОИЗВ. КОМПЕНСАЦИЮ?

$$\tau = -\tau_{max} \operatorname{sign}((E - E_0) \dot{\theta}) \text{ (БЫЛО)}$$

$$\tau_b = -0.8 \tau_{max} \operatorname{sign}((E - E_0) \dot{\theta}) \text{ (СТАЛО)}$$

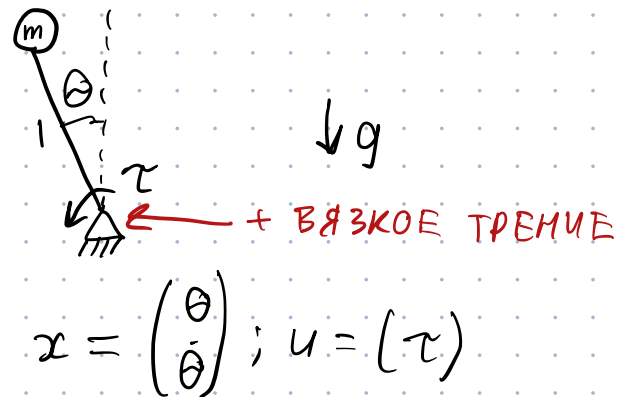
$\tau_{baseline}$ \swarrow compensatory

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

ЕСЛИ μ ИЗВЕСТНО ($\hat{M} = \mu$), ТО

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

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



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

Если $\hat{\mu} \neq \mu$, то

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

КАКИМ СТАНЕТ УР. ДИНАМИКИ ПРИ НЕИДЕАЛЬНОЙ КОМПЕНС.?

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

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

ВВЕДЁМ АДАПТИВН. Ф-Ю УПР. ЛЯПУНОВА:

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

$$E = \frac{ml^2 \dot{\theta}^2}{2} + mgl \cos \theta$$

ЕД. ИЗМ. КОЭФФ. ВЯЗ. ТР
ДЛЯ МАЯТН.:

Н.М.С
РАА

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

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

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

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

\Downarrow

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

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

$$\mu_{\hat{}} - = \mu_{\hat{}} - \dot{} * dt$$

ИДЕНТИФИКАЦИЯ ЛИНЕЙНОЙ СИСТЕМЫ

$$x_{k+1} = A x_k$$

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

$$x_1 = A x_0$$

$$\begin{pmatrix} \dot{p} \\ \ddot{p} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -\frac{\kappa}{m} & 0 \end{pmatrix} \begin{pmatrix} p \\ \dot{p} \end{pmatrix}$$

$$\begin{pmatrix} p_{k+1} \\ \dot{p}_{k+1} \end{pmatrix} = \begin{pmatrix} 1 & \Delta t \\ -\frac{\kappa}{m} \Delta t & 1 \end{pmatrix} \begin{pmatrix} p \\ \dot{p} \end{pmatrix}$$

1)

$$\begin{bmatrix} x_1 & x_2 & \dots & x_N \end{bmatrix} = A \begin{bmatrix} x_0 & x_1 & \dots \end{bmatrix}$$

$X_1 \qquad \qquad X_0$

$$X_1 = A X_0 \quad | \cdot X_0^T$$

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

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

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

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

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

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

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

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

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

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

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

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

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

$$= -2 x_k x_{k+1}^T + 2 A x_k x_k^T$$

$$\hat{A}^- = \lambda \frac{\partial e^T e}{\partial \hat{A}}$$

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