

# Adaptive Estimation of Unknown Inputs with Weakly Nonlinear Dynamics

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# Adaptive Unknown Input Estimators

## Estimator overview

### Three significant uncertainties

- Input  $u$  is unknown, external
- State matrix  $A$  may have uncertainty
- Known, Lipschitz nonlinear internal dynamics  $g(x)$

Can we synthesize  $u$  and correct  $A$ ?

$$\begin{aligned}\dot{x} &= Ax + g(x) + Bu \\ y &= Cx\end{aligned}$$

# Adaptive Unknown Input Estimators

## Modeling unknown inputs

Approximate input space  $\mathbb{U}$

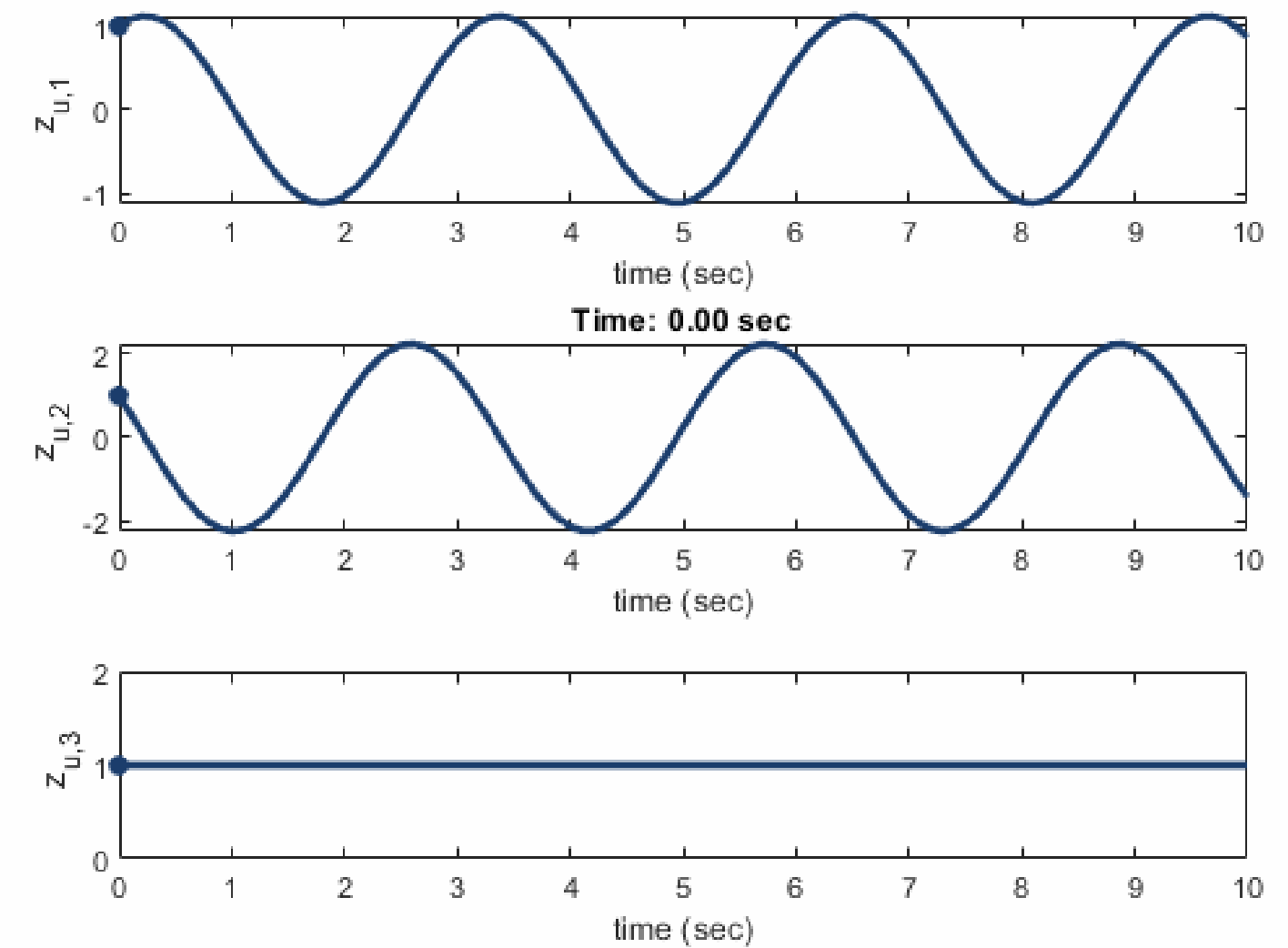
$$- \hat{u} = \sum_{i=1}^N c_i f_i(t)$$

Persisten Inputs

$$- \dot{z}_u = F_u z_u$$

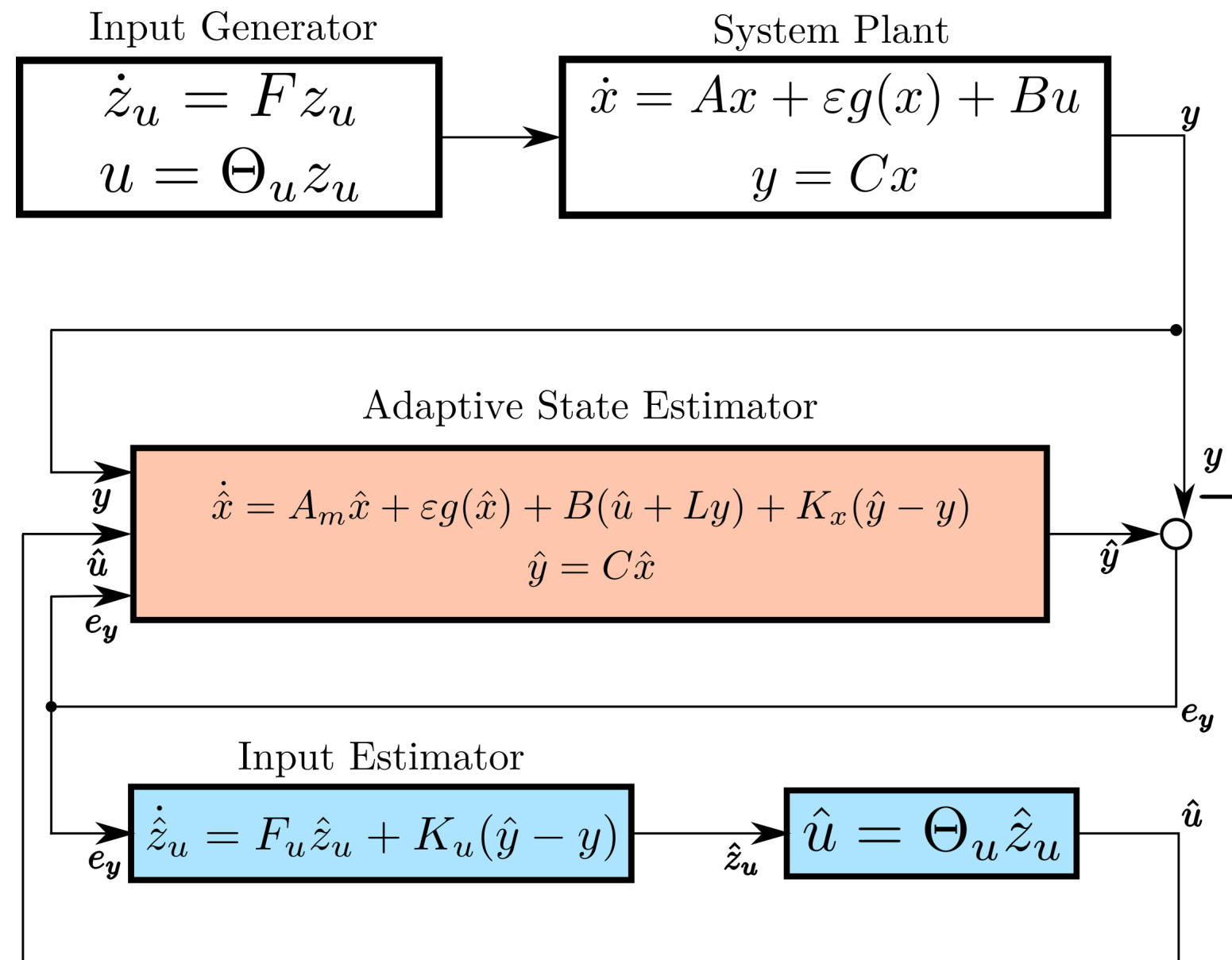
$$- \hat{u} = \Theta_u z_u$$

$$- F_u = \begin{bmatrix} 0 & 1 & 0 \\ -\omega^2 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$



# Adaptive Unknown Input Estimators

## Architecture and estimator error



Recover  $A$  with adaptive scheme

$$A \equiv A_m + BL_*C$$

$$\dot{L} = -e_y y^* \gamma_e; \gamma_e > 0$$

Error dynamics

$$\begin{bmatrix} \dot{e}_x \\ \dot{e}_z \end{bmatrix} = \underbrace{\begin{bmatrix} A_m + K_x C & B\Theta_u \\ K_u C & F \end{bmatrix}}_{\bar{A}_c} \begin{bmatrix} e_x \\ e_z \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} w + \varepsilon \begin{bmatrix} g(\hat{x}) - g(x) \\ 0 \end{bmatrix}$$

# Adaptive Unknown Input Estimators

## Architecture and estimator error

- ASD plant dynamics
- $A$  Hurwitz
- Bounded  $L_*$
- Error in state and input converges to zero
  - $V(e, \Delta L) = \frac{1}{2}e^* \bar{P}e + \frac{1}{2}\text{tr}(\Delta L \gamma_e^{-1} \Delta L^*)$
  - $\dot{V}(e, \Delta L) \leq - \underbrace{\left( \frac{1}{2} \lambda_{\min}(\bar{Q}) - \varepsilon \mu \lambda_{\max}(\bar{P}) \right)}_{\bar{\alpha} > 0} \|e\|^2$

$$0 < \varepsilon < \frac{\lambda_{\min}(\bar{Q})}{2\mu\lambda_{\max}(\bar{P})} \iff \bar{\alpha} > 0.$$

## Illustrative example

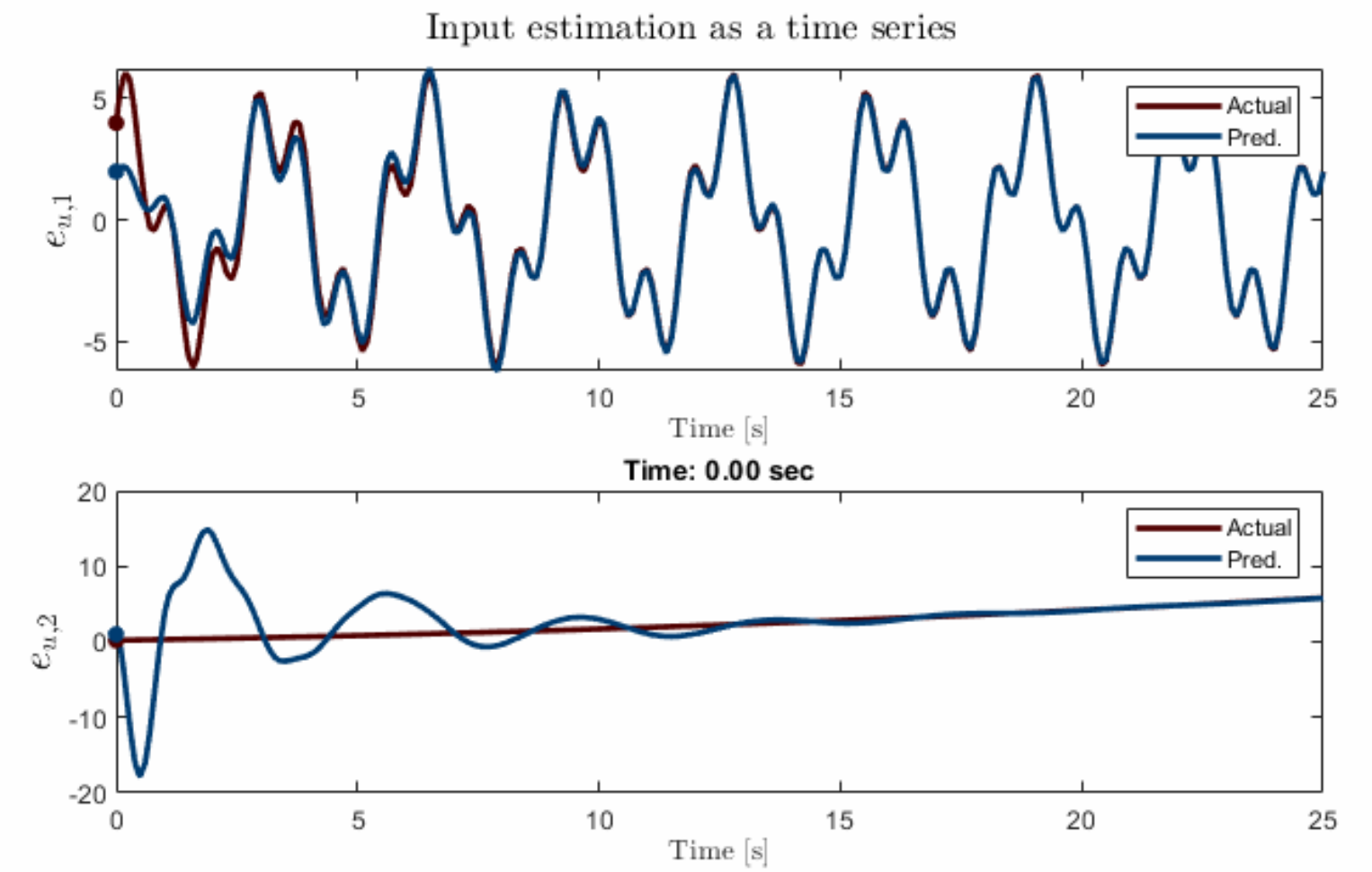
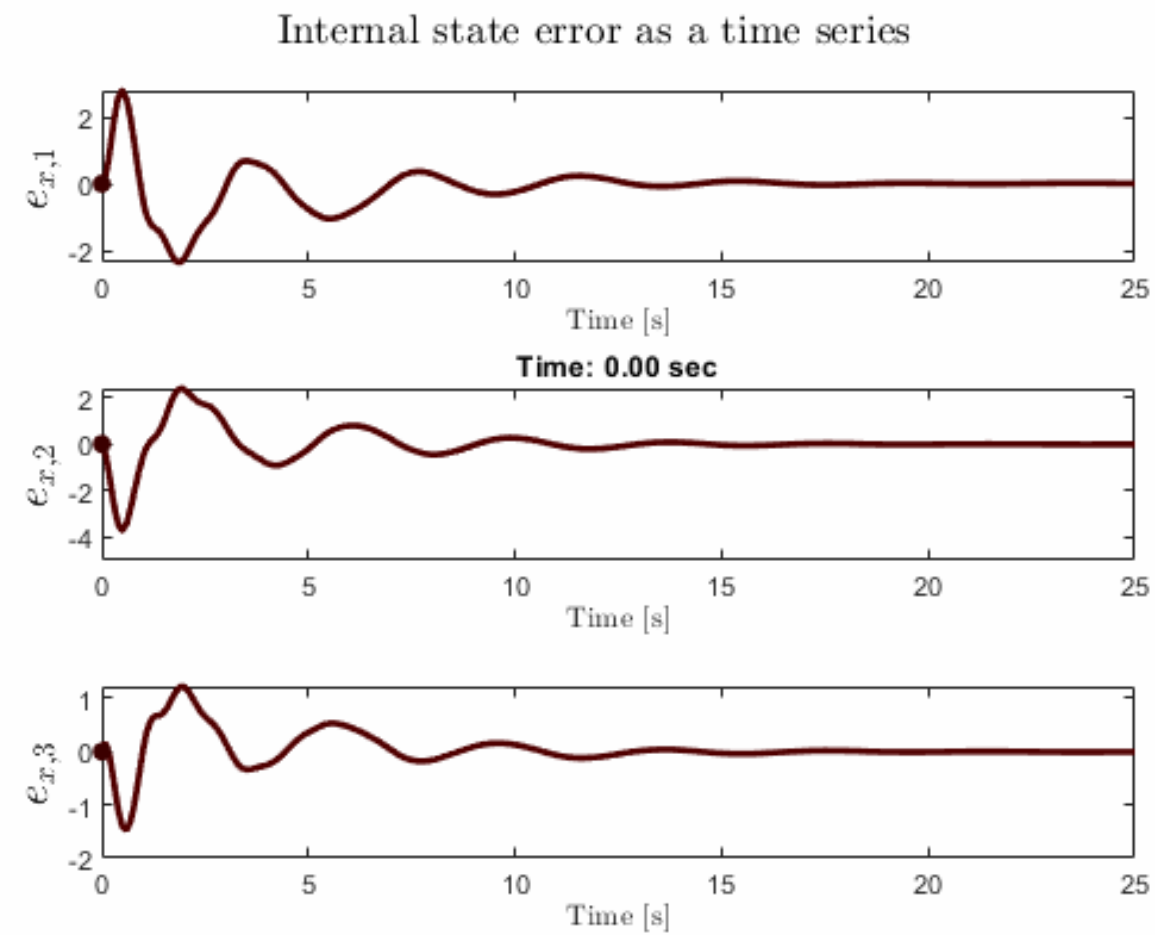
$$\begin{aligned}\dot{x} &= A_m x + \varepsilon g(x) + Bu \\ &= \begin{bmatrix} -4 & 1 & 2 \\ -1 & -1 & 1 \\ -1 & 1 & -1 \end{bmatrix} x + \sin(x) + Bu \\ y &= Cx\end{aligned}$$

$$\begin{aligned}\dot{x} &= Ax + \varepsilon g(x) + Bu \\ &= \begin{bmatrix} -2.86 & 1 & 4.7 \\ 1.8 & -1 & 6.7 \\ -9 & 1 & -17.2 \end{bmatrix} x + \sin(x) + Bu \\ y &= Cx\end{aligned}$$

$$\begin{aligned}L^* &= \begin{bmatrix} -8 & 1 \\ 2 & -7 \end{bmatrix} \\ u_1(t) &= c_{11} \sin(2t) + c_{12} \cos(2t) + c_{13} \sin(7t) + c_{14} \cos(7t) \\ u_2(t) &= c_{11} + c_{22}t + c_{23}t^2 + c_{24}t^3\end{aligned}$$

# Illustrative example

Both the state error and the input error converge simultaneously

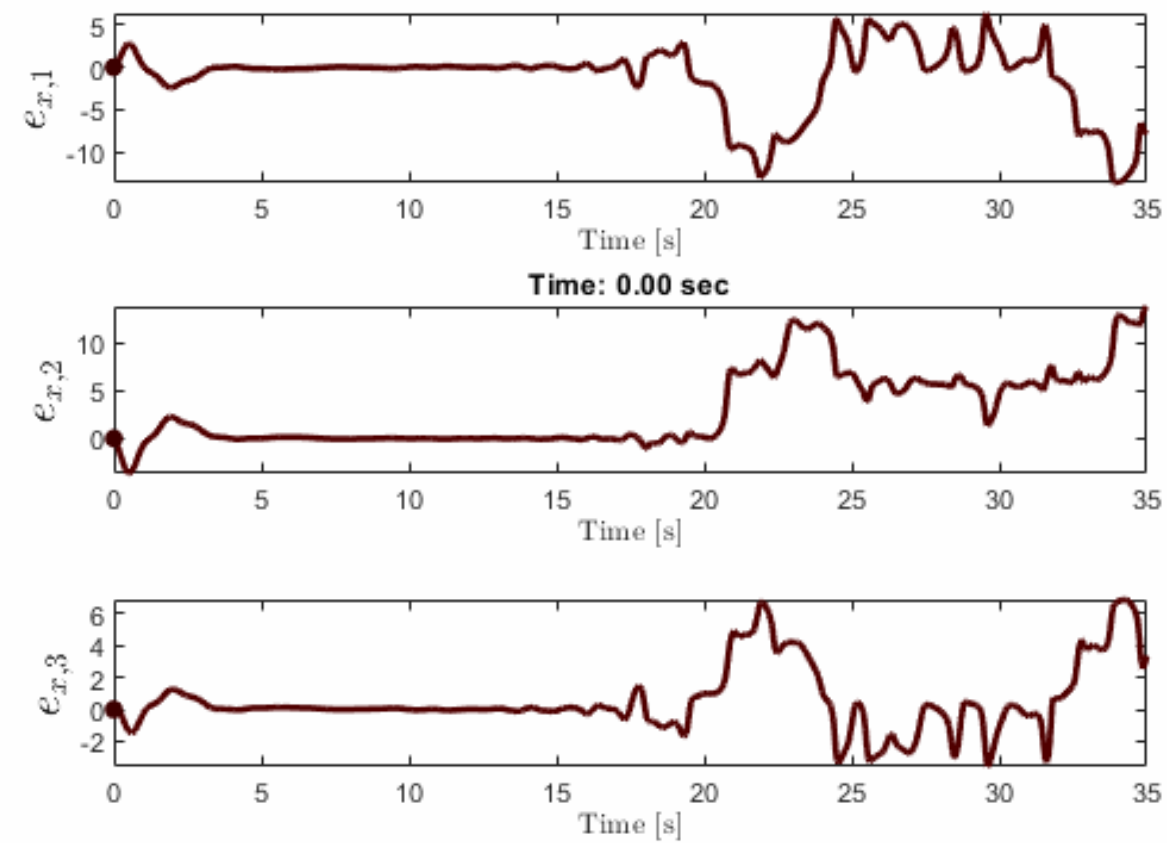


# Illustrative example

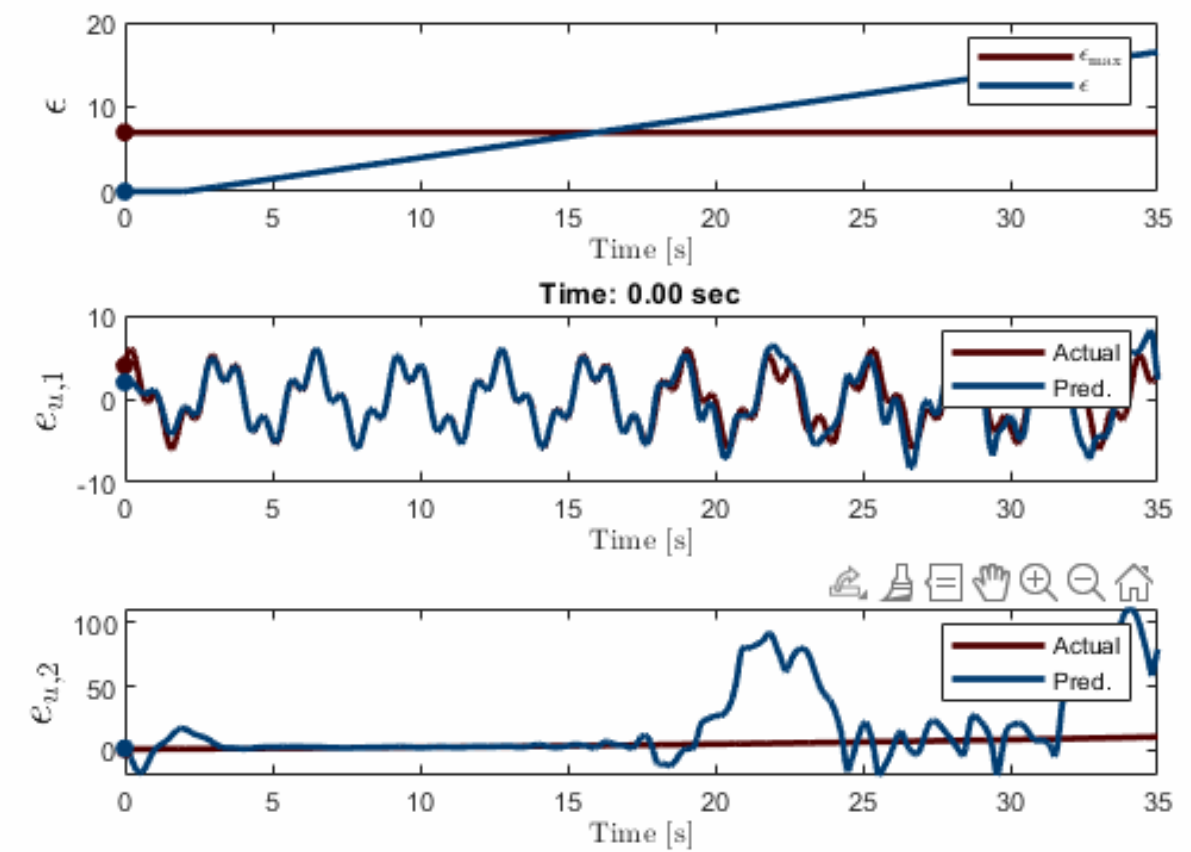
provided  $\epsilon$  is not too great

$$0 < \epsilon < \frac{\lambda_{\min}(\bar{Q})}{2\mu\lambda_{\max}(\bar{P})}$$

Internal state error as a time series



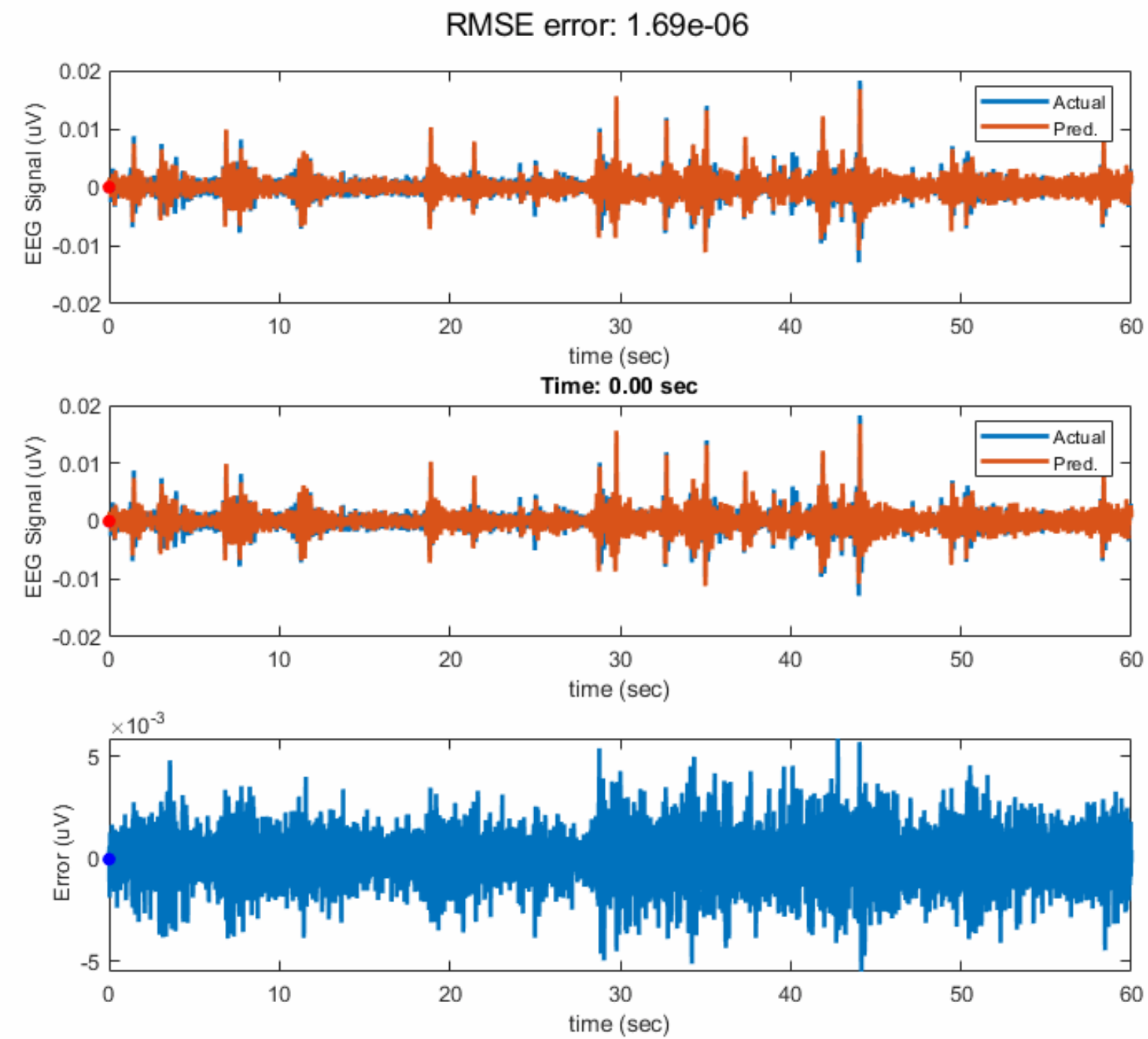
Input estimation as a time series





# Application: Biomarker dynamics

Kalman filtering



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