

Fault detection problem for discrete-time impulsive system using mixed dissipativity approach

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- 1 Introduction
- 2 Problem Formulations
- 3 Mixed Dissipativity Analysis For Fault Diagnosis
 - Mixed dissipativity based system fault sensitivity design
 - Mixed dissipativity based system disturbance insensitivity design
 - Simultaneous mixed fault sensitivity and disturbances insensitivity design
- 4 Case Studies



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Introduction

▶▶ the stability issue

- 📄 Dashkovskiy S, Mironchenko A. “**Input-to-state stability** of nonlinear impulsive systems” *Nonlinear Anal Hybrid Syst.*, vol. 51, no. 3, pp. 1962-1987, 2013.
- 📄 Liu B, Hill DJ, Sun Z. “Mixed \mathcal{K} -dissipativity and stabilization to **ISS** for impulsive hybrid systems” *IEEE Trans Circuits Syst-Exp Briefs.*, vol. 62, no. 8, pp. 791-795, 2015.

▶▶ the filtering problem

- 📄 Pan S, Sun J, Zhao S. “Roust **filtering** for discrete time piecewise impulsive systems” *Signal Process.*, vol. 90, no. 1, pp.324-330, 2010.
- 📄 Xu J, Sun J. “Finite-time **filtering** for discrete time linear impulsive systems” *Signal Process.*, vol. 92, no. 11, pp. 2718-2722, 2012.



Introduction

② the **fault detection** problem

📄 Li W, Yan Y, Bao J. Dissipativity-based distributed fault diagnosis for plantwide chemical process. *J Process Control*. 2020; 96: 37-48.

📄 Zhong M, Xue T, Song Y, Ding SX, Ding EL. Parity space vector machine approach to robust fault detection for linear discrete-time systems. *IEEE Trans Syst Man Cybern Syst*. 2021; 51(7): 4251-4261.



the results do not consider the **impulsive phenomena**.



★ deal with **fault detection** issue of linear discrete-time **impulsive systems**.



Method

In this article, we study the **mixed dissipativity** based fault detection problem for a class of **discrete-time impulsive systems**.

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Dissipativity-based distributed fault diagnosis for plantwide chemical processes[☆]



- ▶ a novel **mixed fault sensitivity** and **mixed disturbance insensitivity** condition;
- ▶ mixed fault sensitivity + mixed disturbance insensitivity $\xrightarrow{\text{mixed supply rate}}$ **mixed dissipativity** condition;
- ▶ the novel mixed dissipativity based fault detection approach is developed for discrete-time impulsive systems.



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System model

$$\begin{cases} x_{k+1} = Ax_k + Df_k + E\omega_k, & k \neq k_m, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m} + D_{\mathcal{I}}f_{k_m} + E_{\mathcal{I}}\omega_{k_m}, & k = k_m, \\ y_k = Cx_k + Wf_k, & k \neq k_m, \\ y_{k_m} = C_{\mathcal{I}}x_{k_m} + W_{\mathcal{I}}f_{k_m}, & k = k_m, \end{cases} \quad (1)$$

- ▶ x_k, y_k : state vector, system output;
- ▶ f_k : **fault signal to be detected**;
- ▶ ω_k : external disturbance;
- ▶ $A, C, D, E, W, A_{\mathcal{I}}, C_{\mathcal{I}}, D_{\mathcal{I}}, E_{\mathcal{I}}, W_{\mathcal{I}}$: known matrices with appropriate dimensions.

⇒

\mathcal{I} : impulse



System model

$$\begin{cases} x_{k+1} = Ax_k + Df_k + E\omega_k, & k \neq k_m, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m} + D_{\mathcal{I}}f_{k_m} + E_{\mathcal{I}}\omega_{k_m}, & k = k_m, \\ y_k = Cx_k + Wf_k, & k \neq k_m, \\ y_{k_m} = C_{\mathcal{I}}x_{k_m} + W_{\mathcal{I}}f_{k_m}, & k = k_m, \end{cases}$$

- ▶ $\{k_m\}_{m \in \mathbb{N}}$: impulsive sequence ($0 = k_0 < k_1 < \dots < k_m < \dots$)
- ▶ $\tau_m =: k_{m+1} - k_m$: impulsive interval

$$0 < \tau_m < \tau < +\infty$$

where τ is the positive constant which represents the **maximum impulsive interval**.



Aims

Construct a residual-based fault diagnoser such that

- 1) sensitive to the fault signal (fault sensitivity)
- 2) insensitive to the exogenous disturbance (disturbance insensitivity)

- ① **fault sensitivity**: the diagnostics unit can quickly and accurately identify fault signals. (**high accuracy**)
- ② **disturbance insensitivity**: the diagnostic device can effectively filter out interference signals that are not related to the fault and avoid false alarms. (**high stability**)



Aims

3) fault sensitivity + disturbance insensitivity conditions
mixed supply rate \rightarrow **mixed dissipativity** conditions

- ▶ **mixed supply rate:** A function $(\gamma_c(u_c, y_c), \gamma_d(u_d, y_d))$ is called a mixed supply rate of impulse system if γ_c is locally integrable and γ_d is locally summable



★ **Fault Diagnoser Design**



Fault diagnoser design

- ▶ First, a fault diagnoser can be designed as follows:

$$\begin{cases} x_{k+1} = Ax_k + Df_k + E\omega_k, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m} + D_{\mathcal{I}}f_{k_m} + E_{\mathcal{I}}\omega_{k_m}, \\ y_k = Cx_k + Wf_k, \\ y_{k_m} = C_{\mathcal{I}}x_{k_m} + W_{\mathcal{I}}f_{k_m}, \end{cases} \Rightarrow$$

$$\begin{cases} \hat{x}_{k+1} = A\hat{x}_k + L(y_k - \hat{y}_k), \\ \hat{x}_{k_m+1} = A_{\mathcal{I}}\hat{x}_{k_m} + L_{\mathcal{I}}(y_{k_m} - \hat{y}_{k_m}), \\ \hat{y}_k = C\hat{x}_k, \\ \hat{y}_{k_m} = C_{\mathcal{I}}\hat{x}_{k_m}, \end{cases}$$

- ▶ \hat{y}_k, \hat{y}_{k_m} : the estimates of y_k and y_{k_m} ;
- ▶ $L, L_{\mathcal{I}}$: the gains to be designed.
- ▶ Second, the residuals are defined by

$$\begin{cases} r_k = M(y_k - \hat{y}_k), & k \neq k_m, \\ r_{k_m} = M_{\mathcal{I}}(y_{k_m} - \hat{y}_{k_m}), & k = k_m, \end{cases}$$

- ▶ $M, M_{\mathcal{I}}$: the residual gain matrices to be designed as well.

\Rightarrow

★ determine the matrices $L, L_{\mathcal{I}}, M, M_{\mathcal{I}}$.



Impulsive error dynamics

Denote estimation error as $e_k = x_k - \hat{x}_k$ and $e_{k_m} = x_{k_m} - \hat{x}_{k_m}$.

$$\begin{cases} x_{k+1} = Ax_k + Df_k + E\omega_k, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m} + D_{\mathcal{I}}f_{k_m} + E_{\mathcal{I}}\omega_{k_m}, \\ y_k = Cx_k + Wf_k, \\ y_{k_m} = C_{\mathcal{I}}x_{k_m} + W_{\mathcal{I}}f_{k_m}, \end{cases} \quad \begin{cases} \hat{x}_{k+1} = A\hat{x}_k + L(y_k - \hat{y}_k), \\ \hat{x}_{k_m+1} = A_{\mathcal{I}}\hat{x}_{k_m} + L_{\mathcal{I}}(y_{k_m} - \hat{y}_{k_m}), \\ \hat{y}_k = C\hat{x}_k, \\ \hat{y}_{k_m} = C_{\mathcal{I}}\hat{x}_{k_m}, \end{cases}$$

$$\Rightarrow \begin{cases} e_{k+1} = Ae_k + E\omega_k + Df_k - L(y_k - \hat{y}_k), & k \neq k_m, \\ e_{k_m+1} = A_{\mathcal{I}}e_{k_m} + E_{\mathcal{I}}\omega_{k_m} + D_{\mathcal{I}}f_{k_m} - L_{\mathcal{I}}(y_{k_m} - \hat{y}_{k_m}), & k = k_m. \end{cases}$$



Impulsive error dynamics

$$\begin{cases} e_{k+1} = Ae_k + E\omega_k + Df_k - L(y_k - \hat{y}_k), & k \neq k_m, \\ e_{k_m+1} = A_{\mathcal{I}}e_{k_m} + E_{\mathcal{I}}\omega_{k_m} + D_{\mathcal{I}}f_{k_m} - L_{\mathcal{I}}(y_{k_m} - \hat{y}_{k_m}), & k = k_m. \end{cases}$$

$$\Rightarrow \begin{cases} e_{k+1} = \tilde{A}e_k + E\omega_k + \tilde{M}f_k, & k \neq k_m, \\ e_{k_m+1} = \tilde{A}_{\mathcal{I}}e_{k_m} + E_{\mathcal{I}}\omega_{k_m} + \tilde{M}_{\mathcal{I}}f_{k_m}, & k = k_m. \end{cases} \quad (2)$$

$$\tilde{A} = A - LC, \tilde{A}_{\mathcal{I}} = A_{\mathcal{I}} - L_{\mathcal{I}}C_{\mathcal{I}}, \tilde{M} = D - LW, \tilde{M}_{\mathcal{I}} = D_{\mathcal{I}} - L_{\mathcal{I}}W_{\mathcal{I}}.$$

$$\begin{cases} r_k = M(y_k - \hat{y}_k) \\ r_{k_m} = M_{\mathcal{I}}(y_{k_m} - \hat{y}_{k_m}) \end{cases} \Rightarrow \begin{cases} r_k = MCe_k + M Wf_k, & k \neq k_m, \\ r_{k_m} = M_{\mathcal{I}}C_{\mathcal{I}}e_{k_m} + M_{\mathcal{I}}W_{\mathcal{I}}f_{k_m}, & k = k_m. \end{cases}$$



Aims

fault sensitivity + disturbance insensitivity \Rightarrow ❓ Specific mathematical form

- ❶ Under zero initial condition and $\omega_k = 0$, the effect of fault f_k on the residual r_k and r_{k_m} should be **sufficiently large**, that is:

$$\sum_{k=k_0+1, k \neq k_m}^T \|r_k\|^2 + \sum_{l=0}^m \|r_{k_l}\|^2 \geq \beta^2 \sum_{k=k_0+1, k \neq k_m}^T \|f_k\|^2 + \beta^2 \sum_{l=0}^m \|f_{k_l}\|^2.$$

- ❷ Under zero initial condition and $f_k = 0$, the effect of exogenous disturbance on the residual should be **minimized**, that is:

$$\sum_{k=k_0+1, k \neq k_m}^T \|r_k\|^2 + \sum_{l=0}^m \|r_{k_l}\|^2 \leq \gamma^2 \sum_{k=k_0+1, k \neq k_m}^T \|\omega_k\|^2 + \gamma^2 \sum_{l=0}^m \|\omega_{k_l}\|^2.$$



Threshold

A fault isolation threshold J_{th} will be designed as:

$$J_r = \left(\sum_{s=k-h, s \neq k_s}^h r_s^\top r_s + \sum_{k_s \in [k-h, h]} r_{k_s}^\top r_{k_s} \right)^{\frac{1}{2}},$$

$$J_{th} = \sup_{\omega \in \mathcal{L}_e, f=0} J_r,$$

$\mathcal{L}_2([0, \infty), \mathbb{R}_n)$ is the space of square summable n -dimensional vector-valued functions. The set \mathcal{L}_{2e} denotes the extended set of \mathcal{L}_2 which consists of the functions whose time truncation lies in \mathcal{L}_2 .

W. Li, Y. Yan, and J. Bao, “Dissipativity-based distributed fault diagnosis for plantwide chemical processes,” *Journal of Process Control*, vol. 96, pp. 37–48, Dec. 2020.

l_2 space

A sequence $\{w_k\}$ belongs to the l_2 space if the sum of its squared elements is finite:

$$\sum_{k=0}^{\infty} |w_k|^2 < \infty$$

$\Rightarrow w_k$ must be bounded and decay over infinite time.

l_{2e} space

A sequence $\{w_k\}$ belongs to the l_{2e} space if it belongs to the l_2 space over some finite time interval. That is, there exists a finite time T such that:

$$\sum_{k=0}^T |w_k|^2 < \infty$$

$\Rightarrow l_{2e}$ allows for signals that are energy-limited over finite periods but not necessarily over infinite periods.

In practical systems, external disturbances are often unpredictable, uncertain,
and may persist indefinitely. $\Rightarrow l_2$ ❌ l_{2e} ✅



Diagnosis

Based on the determined threshold J_{th} , the faults can be detected by comparing J_r with J_{th} according to the following rule:

$$\begin{cases} J_r > J_{th} & \Rightarrow \text{with fault} \Rightarrow \text{alarm,} \\ J_r \leq J_{th} & \Rightarrow \text{no faults.} \end{cases}$$



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Mixed Dissipativity Analysis For Fault Diagnosis

Mixed dissipative

An impulsive system Σ with input $u_k \in \mathcal{U}_c$, $u_{k_m} \in \mathcal{U}_d$, and output $y_k \in \mathcal{Y}_c$, $y_{k_m} \in \mathcal{Y}_d$ is *mixed dissipative* under the mixed supply rate (S_c, S_d) , if there exist storage functions $V_c(x_k, k)$ and $V_d(x_k, k)$, $\forall k$ such that

$$\begin{aligned} \Delta V_c(x_k, k) &< S_c(y_k, u_k), \quad k \neq k_m, \quad \forall u_k \in \mathcal{U}_c, \\ \Delta V_d(x_k, k) &< S_d(y_{k_m}, u_{k_m}), \quad k = k_m, \quad \forall u_{k_m} \in \mathcal{U}_d, \end{aligned}$$

where $\Delta V(x_k, k) = V(x_{k+1}, k+1) - V(x_k, k)$.

1 fault sensitive case

the error dynamics without disturbances is mixed dissipative with the mixed supply rate (S_c, S_d) , which defined as $S_c(f_k, r_k) = \|r_k\|^2 - \beta^2 \|f_k\|^2$ and $S_d(f_{k_m}, r_{k_m}) = \|r_{k_m}\|^2 - \beta^2 \|f_{k_m}\|^2$.

2 disturbance insensitive case

the mixed supply rates are designed as $S_c(\omega_k, r_k) = \gamma^2 \|\omega_k\|^2 - \|r_k\|^2$ and $S_d(\omega_{k_m}, r_{k_m}) = \gamma^2 \|\omega_{k_m}\|^2 - \|r_{k_m}\|^2$

- **?** There is no u in the model.

$$\text{mix dissipative } \begin{cases} \Delta V_c(x_k, k) < S_c(y_k, u_k), \\ \Delta V_d(x_k, k) < S_d(y_{k_m}, u_{k_m}), \end{cases}$$

- **?** Why not y_k and u_k .

$$S_c(f_k, r_k) = \|r_k\|^2 - \beta^2 \|f_k\|^2$$

$$S_d(f_{k_m}, r_{k_m}) = \|r_{k_m}\|^2 - \beta^2 \|f_{k_m}\|^2$$

Note: u_k represents exogenous inputs. Here, it represents either faults f_k or disturbances ω_k .



Mixed dissipativity based system fault sensitivity design

Theorem 1

The error dynamics (2) without disturbances is *mixed dissipative* under the mixed supply rate (S_c, S_d) with $S_c(f_k, r_k) = \|r_k\|^2 - \beta^2 \|f_k\|^2$ and $S_d(f_{k_m}, r_{k_m}) = \|r_{k_m}\|^2 - \beta^2 \|f_{k_m}\|^2$, if there exist matrices Y, Y_I , positive semi-definite matrices Z, Z_I , a positive definite matrix P , scalar β such that the following LMIs hold:

$$\Pi_1 = \begin{bmatrix} P & PA - YC & PD - YW \\ & P + C^\top ZC & C^\top ZW \\ & * & W^\top ZW - \beta^2 I \end{bmatrix} > 0, \quad (3)$$

$$\Pi_2 = \begin{bmatrix} P & PA_I - Y_I C_I & PD_I - Y_I W_I \\ & P + C_I^\top Z_I C_I & C_I^\top Z_I W_I \\ & * & W_I^\top Z_I W_I - \beta^2 I \end{bmatrix} > 0. \quad (4)$$



Mixed dissipativity based system fault sensitivity design

In this case, the gains of the desired diagnoser can be obtained as follows:

$$L = P^{-1}Y, \quad L_I = P^{-1}Y_I, \quad Z = M^\top M, \quad Z_I = M_I^\top M_I$$

The residual gains M and M_I can be obtained by factorizing Z, Z_I .

证明.

- ▶ **Part 1.** Mixed dissipative.
- ▶ **Part 2.** Mixed fault sensitivity.

Omitted here. □



Mixed dissipativity based system disturbance insensitivity design

Theorem 2

The impulsive error dynamics (2) without faults is *mixed dissipative* under the mixed supply rate (S_c, S_d) with $S_c(\omega_k, r_k) = \gamma^2 \|\omega_k\|^2 - \|r_k\|^2$ and $S_d(\omega_{k_m}, r_{k_m}) = \gamma^2 \|\omega_{k_m}\|^2 - \|r_{k_m}\|^2$, if there exist matrices Y, Y_I , positive semi-definite matrices Z, Z_I , positive definite matrix P , scalar γ such that the following LMIs hold:

$$\Gamma_1 = \begin{bmatrix} P & PA - YC & PE \\ & P - C^\top ZC & 0 \\ & * & \gamma^2 I \end{bmatrix} > 0, \quad (5)$$

$$\Gamma_2 = \begin{bmatrix} P & PA_I - Y_I C_I & PE_I \\ & P - C_I^\top Z_I C_I & 0 \\ & * & \gamma^2 I \end{bmatrix} > 0. \quad (6)$$



Mixed dissipativity based system disturbance insensitivity design

In this case, the gains of the desired fault diagnostic observer can be obtained as follows:

$$L = P^{-1} Y, \quad L_I = P^{-1} Y_I, \quad Z = M^\top M, \quad Z_I = M_I^\top M_I$$

The residual gains M and M_I can be obtained by factorizing Z, Z_I .

证明.

- ▶ **Part 1.** Mixed dissipative.
- ▶ **Part 2.** Mixed disturbances insensitivity.

Omitted here. □



Simultaneous mixed fault sensitivity and disturbances insensitivity design

Theorem 3

The impulsive error dynamics (2) satisfies the fault sensitivity, disturbances insensitivity conditions simultaneously, if there exist matrices Y, Y_L , positive semi-definite matrices Z, Z_L , a positive definite matrix P , and scalars β, γ such that the following LMIs hold:

$$\begin{aligned} \max_{\beta, \gamma} \quad & \beta - \gamma \\ \text{s.t.} \quad & \Pi_1 > 0, \quad \Pi_2 > 0 \quad \Gamma_1 > 0, \quad \Gamma_2 > 0. \end{aligned} \quad (7)$$



Simultaneous mixed fault sensitivity and disturbances insensitivity design

In this case, the gains of the desired fault diagnostic observer can be obtained as follows:

$$L = P^{-1} Y, \quad L_{\mathcal{I}} = P^{-1} Y_{\mathcal{I}}, \quad Z = M^{\top} M, \quad Z_{\mathcal{I}} = M_{\mathcal{I}}^{\top} M_{\mathcal{I}}$$

The residual gains M and $M_{\mathcal{I}}$ can be obtained by factorizing Z , $Z_{\mathcal{I}}$.

证明.

The proof is straightforward with the combination of Theorems 1 and 2. □



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Parameters

$$\begin{cases} x_{k+1} = Ax_k + Df_k + E\omega_k, & k \neq k_m, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m} + D_{\mathcal{I}}f_{k_m} + E_{\mathcal{I}}\omega_{k_m}, & k = k_m, \\ y_k = Cx_k + Wf_k, & k \neq k_m, \\ y_{k_m} = C_{\mathcal{I}}x_{k_m} + W_{\mathcal{I}}f_{k_m}, & k = k_m, \end{cases}$$

► coefficient matrix

$$\begin{aligned} A &= \begin{bmatrix} 0.3658 & 0.16 \\ 0.6532 & 0.6169 \end{bmatrix}, A_{\mathcal{I}} = \begin{bmatrix} 0.2 & 0 \\ 0 & 0.2 \end{bmatrix}, E_{\mathcal{I}} = \begin{bmatrix} -0.1 \\ 0.7 \end{bmatrix}, W = 1, \\ D &= \begin{bmatrix} 0.8 \\ -1 \end{bmatrix}, D_{\mathcal{I}} = \begin{bmatrix} 0.7 \\ -1 \end{bmatrix}, C = \begin{bmatrix} 0 \\ 0.1 \end{bmatrix}, C_{\mathcal{I}} = \begin{bmatrix} 0 \\ 0.2 \end{bmatrix}, E = \begin{bmatrix} -0.1 \\ 0.7 \end{bmatrix}, W_{\mathcal{I}} = 1. \end{aligned}$$



Gains

- ▶ impulse sequence、 interval

$$k_m = 5l \ (l = 0, 1, 2, \dots) \Rightarrow \tau_m = 5$$

- ▶ Calculate the LMIs (3), (4) and (5), (6) obtained in Theorems 1 and 2, respectively.

Table 1: Observer gains and other calculated variables

Variables	β	γ	M	$M_{\mathcal{I}}$	P		L	$L_{\mathcal{I}}$
Values	2.225	0.7136	4.5115	3.9875	1.4036	0.2736	0.3768	0.3854
					0.2736	0.3453	2.843	2.7583



Noise/Fault

- ▶ The system external noise

$$\omega_k = 0.4 \sin\left(\frac{\pi}{4}k\right).$$

- ▶ The fault signal

$$f(x) = \begin{cases} 0.1, & 20 \leq k \leq 30, \\ 0, & \text{otherwise.} \end{cases}$$



Evaluation of $J(k)$

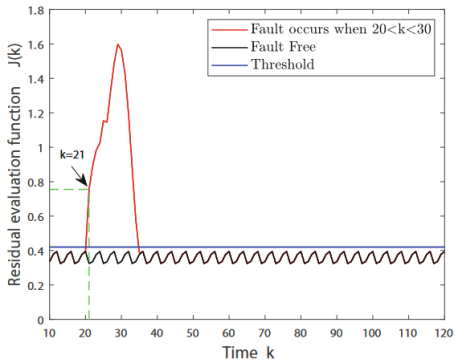


Figure 1: Residual evaluation function $J(k)$



Mass-spring system

$$m_1 \ddot{q}_1(t) + (k_1 + k_2)q_1(t) - k_2 q_2(t) = F,$$

$$m_2 \ddot{q}_2(t) - k_2 q_1(t) + k_2 q_2(t) = 0,$$

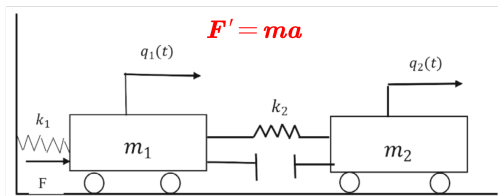
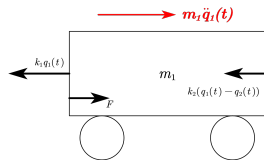
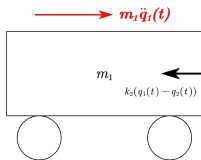
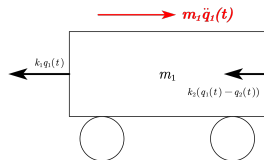
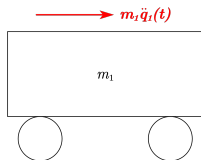
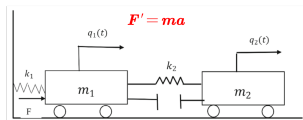


Figure 2: Two-mass systems with constraint buffer



Mass-spring system



$$\Rightarrow m_1 \ddot{q}_1(t) = -k_1 q_1(t) - k_2 (q_1(t) - q_2(t)) + F$$



Parameters

$$\begin{aligned}m_1 \ddot{q}_1(t) + (k_1 + k_2)q_1(t) - k_2 q_2(t) &= F, \\m_2 \ddot{q}_2(t) - k_2 q_1(t) + k_2 q_2(t) &= 0,\end{aligned}$$

- ▶ coefficient

$$m_1 = 1.5\text{kg}, m_2 = 0.5\text{kg}, k_1 = 1\text{N/m}, k_2 = 0.5\text{N/m}.$$

- ▶ F is the external force and represents the **fault signal** in this mass-spring system.

$$F = \begin{cases} 0.088\text{N}, & 50 \leq k \leq 55, \\ 0, & \text{otherwise.} \end{cases}$$



Parameters

- ▶ the loss of kinetic energy due to a collision

$$\begin{aligned}
 m_1 \dot{q}_1(t_m + 1) + m_2 \dot{q}_2(t_m + 1) &= m_1 \dot{q}_1(t_m) + m_2 \dot{q}_2(t_m), \\
 \dot{q}_1(t_m + 1) - \dot{q}_2(t_m + 1) &= -\delta(\dot{q}_1(t_m) - \dot{q}_2(t_m)), \quad (8)
 \end{aligned}$$

where $\delta \in [0, 1)$, is the coefficient restitution.

📖 Haddad WM, Chellaboina V, Nersesov SG. Impulsive and Hybrid Dynamical Systems: Stability, Dissipativity, and Control. *Princeton University Press*; 2006.

- ▶ $x(t) = [q_1(t), \dot{q}_1(t), q_2(t), \dot{q}_2(t)]^\top$
the collision occurs when $x \in \mathcal{J} = \{x_1 - x_3 = l\}$
 - ▶ x_i : the i -th component of the state $x(t)$
 - ▶ l : the total buffer length
- ▶ $y = [x_1^\top, x_3^\top]^\top$



Parameters

$$\Rightarrow \begin{cases} \dot{x}(t) = Ax(t) + DF, & k \neq k_m, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m}, & k = k_m, \end{cases} \quad (9)$$

where

$$A_{\mathcal{I}} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & -\frac{(1+\delta)m_2}{m_1+m_2} & 0 & \frac{(1+\delta)m_2}{m_1+m_2} \\ 0 & 0 & 1 & 0 \\ \frac{(1+\delta)m_1}{m_1+m_2} & 0 & -\frac{(1+\delta)m_1}{m_1+m_2} & 0 \end{bmatrix},$$

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -\frac{k_1+k_2}{m_1} & 0 & \frac{k_2}{m_1} & 0 \\ 0 & 0 & 0 & 1 \\ \frac{k_2}{m_2} & 0 & -\frac{k_2}{m_2} & 0 \end{bmatrix}, \quad D = \begin{bmatrix} 0 \\ \frac{1}{m_1} \\ 0 \\ 0 \end{bmatrix}.$$



Parameters

- ▶ Let the sampling interval $S = 0.5s$

$$\begin{cases} \dot{x}(t) = Ax(t) + DF, & k \neq k_m, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m}, & k = k_m, \end{cases} \Rightarrow \begin{cases} x_{k+1} = \bar{A}x_k + Hf_k, & k \neq k_m, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m}, & k = k_m, \end{cases} \quad (10)$$

where $\bar{A} = e^{AS}$, $H = \left(\int_0^S e^{At} dt \right) D$.

- ▶ Noise

$$\Rightarrow \begin{cases} x_{k+1} = \bar{A}x_k + Hf_k + E\omega_k, & k \neq k_m, \\ x_{k_m+1} = A_{\mathcal{I}}x_{k_m}, & k = k_m, \end{cases} \quad (11)$$

where $\omega_k = \sin\left(\frac{\pi}{4}k\right)$



Gains

By solving the LMI (7), the feasible solutions are obtained in Table 2.

Table 2: Observer gains and other calculated variables

Variables	β	γ	M	L
Values	0.7985	0.5698	1.146	$\begin{bmatrix} -0.0938 \\ -0.4743 \\ 0.8897 \\ 1.6210 \end{bmatrix}$



Evaluation of $J(k)$

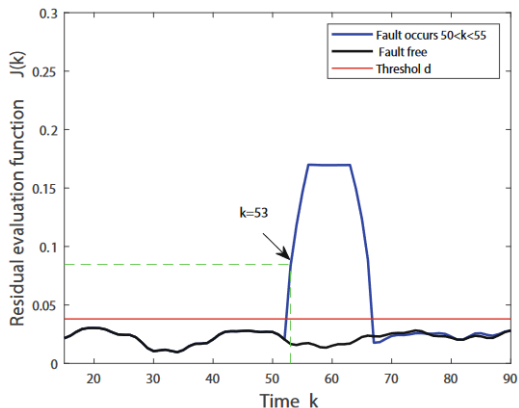


Figure 3: Residual evaluation function $J(k)$



Thank you for your attention!