

Closed loop information processing strategy for optimal fault detection and control

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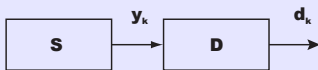
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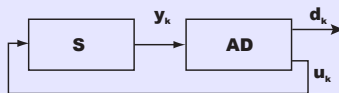
Introduction

Model-based approaches to the fault detection

- Uses a model of the observed system, a priori information and measurements to decide on faults
- Passive fault detection - detector passively uses available information to decide on faults
- Active fault detection - detector provides decision and input signal which should improve fault detection



a) **Passive fault detection**



b) **Active fault detection**

Introduction – cont'd

Active fault detection problem

- Deterministic [Campbell&Nikoukhah(2004)] and stochastic [Zhang(1989), Kerestecioglu(1993)] models of the observed system are used
- A general formulation of the active fault detection problem in stochastic framework is missing and the relation between active fault detection and the optimal control is not considered
- Known approaches use information in such way that the consequences of the current decision in future steps are not considered and the future losses are not taken into account

Introduction – cont'd

Information processing strategies

- Open loop (OL) - only a priori information is used
- Open loop feedback (OLF) - all available information up to current time step is used, but the future information is not considered
- Closed loop (CL) - all available information up to current time step is used and the availability of the future information is considered as well; so the future losses are taken into account and this strategy provides the lowest value of a criterion (i.e. $J^{CL} \leq J^{OLF} \leq J^{OL}$)

Introduction – cont'd

Goals

- Propose a unified formulation of the active fault detection problem
- Specify three basic special cases
 - Optimal detector for given input signal generator
 - Optimal detector and optimal input signal generator
 - Optimal detector and optimal dual controller
- Find solutions of considered special cases using CL information processing strategy

Problem formulation

Description of the observed system for $k \in \mathcal{T} = \{0, \dots, F\}$

$$\text{System : } \mathbf{x}_{k+1} = \mathbf{f}_k(\mathbf{x}_k, \mu_k \mathbf{u}_k, \mathbf{w}_k)$$

$$\mu_{k+1} = \mathbf{g}_k(\mu_k, \mathbf{e}_k)$$

$$\mathbf{y}_k = \mathbf{h}_k(\mathbf{x}_k, \mu_k, \mathbf{v}_k)$$

\mathbf{f}_k , \mathbf{g}_k and \mathbf{h}_k are known functions; $\mathbf{x}_k \in \mathcal{R}^{n_x}$ is controllable part of the state; $\mu_k \in \mathcal{M} \subset \mathcal{R}^{n_\mu}$ is uncontrollable part of the state and represents faults; $\mathbf{u}_k \in \mathcal{U}_k \subset \mathcal{R}^{n_u}$ is input, $\mathbf{y}_k \in \mathcal{R}^{n_y}$ is output; $\{\mathbf{w}_k\}$, $\{\mathbf{e}_k\}$ and $\{\mathbf{v}_k\}$ are mutually independent random sequences

Problem formulation – cont'd

Description of the general active detector for $k \in \mathcal{T}$

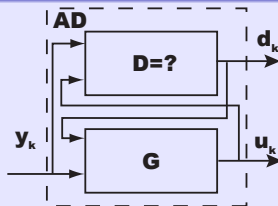
$$\text{Active detector : } \begin{bmatrix} d_k \\ \mathbf{u}_k \end{bmatrix} = \begin{bmatrix} \sigma_k(\mathbf{l}_0^k) \\ \gamma_k(\mathbf{l}_0^k, d_k) \end{bmatrix}$$

σ_k and γ_k are generally unknown functions; $d_k \in \mathcal{M}$ is an estimate of the μ_k (in special case called decision); all available information at time k is stored in $\mathbf{l}_0^k = [\mathbf{y}_0^k{}^T, \mathbf{u}_0^{k-1}{}^T, d_0^{k-1}{}^T]^T$

Specifications and solutions of the special cases

Case I: Optimal detector for given input signal generator

- The input signal generator is given by functions $\gamma_k(\mathbf{I}_0^k, d_k)$ (e.g. existing controller) and the detector $\sigma_k(\mathbf{I}_0^k)$ has to be found



- Criterion which has to be minimized

$$J_{ADGG}(\sigma_0^F) = \mathbb{E} \left\{ \sum_{i=0}^F L_i^d(d_i, \mu_i) \right\}$$

Specifications and solutions of the special cases – cont'd

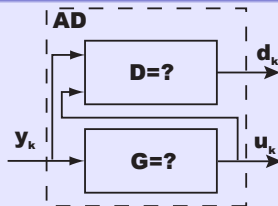
Case I: Optimal detector for given input signal generator

- Backward recursive equation with initial condition $V_{F+1}^* = 0$ and final value of the criterion $J_{ADGG}^{CL} = E \{ V_0^*(\mathbf{I}_0) \}$
$$V_k^*(\mathbf{I}_0^k) = \min_{d_k \in \mathcal{M}} E \left\{ L_k^d(d_k, \mu_k) + V_{k+1}^*(\mathbf{I}_0^{k+1}) | \mathbf{I}_0^k, d_k \right\}$$
- Optimal decision d_k^* is a trade-off between right decision at time step k and the excitation of the observed system through the law $\mathbf{u}_k = \gamma_k(\mathbf{I}_0^k, d_k)$

Specifications and solutions of the special cases – cont'd

Case II: Optimal detector and optimal input signal generator

- Both the detector $\sigma_k(\mathbf{I}_0^k)$ and input signal generator $\gamma_k(\mathbf{I}_0^k, d_k)$ have to be found



- Criterion which has to be minimized

$$J_{ADG}(\sigma_0^F, \gamma_0^F) = \mathbb{E} \left\{ \sum_{i=0}^F L_i^d(d_i, \mu_i) \right\}$$

Specifications and solutions of the special cases – cont'd

Case II: Optimal detector and optimal input signal generator

- Backward recursive equation with the initial condition

$$V_{F+1}^* = 0 \text{ and final value of the criterion } J_{ADG}^{CL} = E \{ V_0^*(\mathbf{y}_0) \}$$

$$V_k^*(\mathbf{y}_0^k, \mathbf{u}_0^{k-1}) = \min_{d_k \in \mathcal{M}} E \left\{ L_k^d(d_k, \mu_k) | \mathbf{y}_0^k, \mathbf{u}_0^{k-1}, d_k \right\} +$$

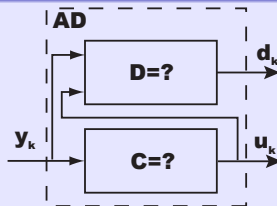
$$\min_{\mathbf{u}_k \in \mathcal{U}_k} E \left\{ V_{k+1}^*(\mathbf{y}_0^{k+1}, \mathbf{u}_0^k) | \mathbf{y}_0^k, \mathbf{u}_0^k \right\}$$

- Optimal decision d_k^* and optimal input signal \mathbf{u}_k^* are chosen independently, so $\gamma_k(\mathbf{y}_0^k, \mathbf{u}_0^{k-1}, d_0^k) = \gamma_k(\mathbf{y}_0^k, \mathbf{u}_0^{k-1})$
- It can be proven that $J_{ADG}^{CL} \leq J_{ADGG}^{CL}$

Specifications and solutions of the special cases – cont'd

Case III: Optimal detector and optimal dual controller

- Both the detector $\sigma_k(\mathbf{I}_0^k)$ and dual controller $\gamma_k(\mathbf{I}_0^k, d_k)$ have to be found



- Criterion which has to be minimized

$$J_{ADC}(\sigma_0^F) = \mathbb{E} \left\{ \sum_{i=0}^F L_i^d(d_i, \mu_i) + \alpha_i L_i^c(\mathbf{x}_i, \mathbf{u}_i) \right\}$$

Specifications and solutions of the special cases – cont'd

Case III: Optimal detector and optimal dual controller

- Backward recursive equation with initial condition $V_{F+1}^* = 0$ and final value of the criterion $J_{ADC}^{CL} = E \{ V_0^*(\mathbf{y}_0) \}$

$$V_k^*(\mathbf{y}_0^k, \mathbf{u}_0^{k-1}) = \min_{d_k \in \mathcal{M}} E \left\{ L_k^d(d_k, \mu_k) | \mathbf{y}_0^k, \mathbf{u}_0^{k-1}, d_k \right\} +$$

$$\min_{\mathbf{u}_k \in \mathcal{U}_k} E \left\{ \alpha_k L_k^c(\mathbf{x}_k, \mathbf{u}_k) + V_{k+1}^*(\mathbf{y}_0^{k+1}, \mathbf{u}_0^k) | \mathbf{y}_0^k, \mathbf{u}_0^k \right\}$$
- Optimal decision d_k^* and optimal input signal \mathbf{u}_k^* are also independent, but optimal input signal is trade-off between control objective and the excitation of the observed system

Specifications and solutions of the special cases – cont'd

Comments on special cases

- The case III: Optimal detector and optimal dual controller describes the most general problem and it includes the previous cases
- A nonlinear filter can provide required pdf's $p(\mu_k | \mathbf{y}_0^k, \mathbf{u}_0^{k-1})$, $p(\mathbf{x}_k | \mathbf{y}_0^k, \mathbf{u}_0^{k-1})$ and $p(\mathbf{y}_{k+1} | \mathbf{y}_0^k, \mathbf{u}_0^k)$
- The function $V_k^*(\cdot)$ can not be expressed analytically in explicit form and the approximations are necessary to obtain a feasible suboptimal solution

Special case – Multiple linear Gaussian model

Specification of the MM framework

- It is special case of the proposed description of the observed system and it simplifies a computation
- The observed system is supposed to be described by finite number of linear Gaussian models, so pdf's of the initial condition \mathbf{x}_k and noises w_k, v_k are Gaussian
- The set $\mathcal{M} = 1, \dots, N$ is discrete and μ_k denotes scalar index to the model valid at time k
- The state equation $\mu_{k+1} = \mathbf{g}(\mu_k, \mathbf{e}_k)$ is replaced by transition probabilities $P_{i,j} = P(\mu_{k+1} = j | \mu_k = i), i, j \in \mathcal{M}$

Numerical example

Observed system description for $k \in \mathcal{T} = \{0, 1\}$

$$\mu_k = 1 : x_{k+1} = 0.99x_k + u_k + \sqrt{0.25}w_k$$

$$y_k = 2x_k + \sqrt{0.25}v_k$$

$$\mu_k = 2 : x_{k+1} = 1.01x_k + 0.99u_k + \sqrt{0.25}w_k$$

$$y_k = 2x_k + \sqrt{0.25}v_k$$

$$\mu_k = 3 : x_{k+1} = 0.5x_k + 1.5u_k + \sqrt{0.25}w_k$$

$$y_k = 1.5x_k + \sqrt{0.25}v_k$$

$$P_{i,j} = \begin{cases} 0.9 & \text{iff } i = j, \\ 0.05 & \text{iff } i \neq j, \end{cases} \quad p(w_k) = p(v_k) = \mathcal{N}\{0, 1\}, \quad p(x_0) = \mathcal{N}\{1, 0.1\}$$
$$P(\mu_0 = 1) = 0.4, \quad P(\mu_0 = 2) = P(\mu_0 = 3) = 0.3$$

Numerical example – cont'd

Loss functions for $k \in \mathcal{T}$

$$\begin{aligned}d_k = \mu_k &\Rightarrow L_k^d(d_k, \mu_k) = 0 & L_k^c(x_k, u_k) &= x_k^2 + u_k^2 \\d_k \neq \mu_k &\Rightarrow L_k^d(d_k, \mu_k) = 1 & \alpha &= 0.01\end{aligned}$$

Input signal generator for $k \in \mathcal{T}$

- Set of input signal $\mathcal{U}_k = \{-1, 1\}$
- Description of the generator for the case I

$$\begin{aligned}d_k = 1 \vee d_k = 3 &\Rightarrow u_k = -1, \\d_k = 2 &\Rightarrow u_k = 1\end{aligned}$$

Numerical example – cont'd

Approaches used for active detector design

- Case I (Optimal detector for given input signal generator) is solved using OLF (MAP model estimate) and CL strategy
- Case II (Optimal detector and optimal input signal generator) is solved only using CL strategy
- Case III (Optimal detector and optimal dual controller) is solved using certainty equivalence (CE) and CL strategy

Results of Monte Carlo simulations

\hat{J}_{ADGG}^{OLF}	\hat{J}_{ADGG}^{CL}	\hat{J}_{ADG}^{CL}	\hat{J}_{ADC}^{CE}	\hat{J}_{ADC}^{CL}
1.2647	1.2323	1.0890	1.3022	1.1655

Concluding remarks

Remarks

- The new unified formulation of the active fault detection problem was proposed
- The formulation provides very general framework and the other cases together with corresponding solutions can be easily derived in addition to the presented cases
- In general, CL information processing strategy provides better results than OLF strategy
- In the case II (Optimal detector and optimal input signal generator) it was shown that the optimal decision d_k^* and optimal input signal \mathbf{u}_k^* are independent