

# Unified solution of optimal active fault detection and optimal control

Miroslav Šimandl   Ivo Punčochář

Department of Cybernetics  
Faculty of Applied Sciences  
University of West Bohemia in Pilsen

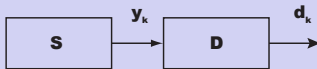
# Outline

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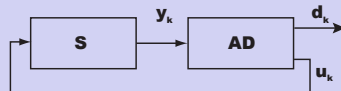
# Introduction

## Passive and active fault detection problem

- Passive fault detection - detector passively uses available information to decide on faults
- Active fault detection - active detector provides decision and input signal that 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 active fault detection problem
- Solve given problem using CL information processing strategy
- Focus on design of active detector containing dual controller
- Discuss two interesting special cases of active fault detection

## Problem formulation

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

$$\mathbf{x}_{k+1} = \mathbf{f}_k(\mathbf{x}_k, \boldsymbol{\mu}_k \mathbf{u}_k, \mathbf{w}_k)$$

$$\boldsymbol{\mu}_{k+1} = \mathbf{g}_k(\boldsymbol{\mu}_k, \mathbf{e}_k)$$

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

$\mathbf{f}_k$ ,  $\mathbf{g}_k$  and  $\mathbf{h}_k$  are known function;  $\mathbf{x}_k \in \mathcal{R}^{n_x}$  is controllable part of the state;  $\boldsymbol{\mu}_k \in \mathcal{M} \subset \mathcal{R}^{n_\mu}$  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 white and mutually independent random sequences

## Problem formulation – cont'd

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

$$\begin{bmatrix} \mathbf{d}_k \\ \mathbf{u}_k \end{bmatrix} = \rho_k \left( \mathbf{l}_0^k \right) = \begin{bmatrix} \sigma_k \left( \mathbf{l}_0^k \right) \\ \gamma_k \left( \mathbf{l}_0^k, \mathbf{d}_k \right) \end{bmatrix}$$

Functions  $\rho_k$  are unknown and should be designed. Both decision  $\mathbf{d}_k$  and input  $\mathbf{u}_k$  are generated by the active detector. Information vector  $\mathbf{l}_0^k = \left[ \mathbf{y}_0^k{}^T, \mathbf{u}_0^{k-1}{}^T, \mathbf{d}_0^{k-1}{}^T \right]^T$  contains all information received up to time step  $k$ .



## Problem formulation – cont'd

### Design criterion for active detector containing dual controller

$$J(\rho_0^F) = \mathbb{E} \left\{ \sum_{i=0}^F L_k^d(\mathbf{d}_k, \boldsymbol{\mu}_k) + \alpha_k L_k^c(\mathbf{x}_k, \mathbf{u}_k) \right\}$$

The cost function  $L_k^d$  penalizes a difference between  $\mathbf{d}_k$  and  $\boldsymbol{\mu}_k$ . The cost function  $L_k^c$  penalizes  $\mathbf{x}_k$  and  $\mathbf{u}_k$ . The coefficient  $\alpha_k \geq 0$  sets a desired compromise between active detection and dual control objectives.

## Unified solution

### Solution of dynamic optimization problem

- Minimization of the criterion  $J(\rho_0^F)$  is solved using the backward recursive equation (BRE)

$$V_k^* \left( \mathbf{l}_0^k \right) = \min_{\substack{\mathbf{d}_k \in \mathcal{M} \\ \mathbf{u}_k \in \mathcal{U}_k}} \mathbb{E} \left\{ L_k^d(\mathbf{d}_k, \boldsymbol{\mu}_k) + \alpha_k L_k^c(\mathbf{x}_k, \mathbf{u}_k) \right. \\ \left. + V_{k+1}^* \left( \mathbf{l}_0^{k+1} \right) \mid \mathbf{l}_0^k, \mathbf{u}_k, \mathbf{d}_k \right\}$$

with initial condition  $V_{F+1}^* = 0$  and minimal value of the criterion  $J$  is  $J^* = \mathbb{E} \{ V_0(\mathbf{l}_0) \}$

- The filtering pdf  $p(\mathbf{x}_k, \boldsymbol{\mu}_k \mid \mathbf{y}_0^k, \mathbf{u}_0^k, \mathbf{d}_0^k)$  and the predictive pdf  $p(\mathbf{y}_{k+1} \mid \mathbf{y}_0^k, \mathbf{u}_0^k, \mathbf{d}_0^k)$  are required to solve BRE

## Unified solution – cont'd

### Filtering and predictive pdf's

- It can be shown the following identities hold

$$p(\mathbf{x}_k, \boldsymbol{\mu}_k | \mathbf{y}_0^k, \mathbf{u}_0^k, \mathbf{d}_0^k) = p(\mathbf{x}_k, \boldsymbol{\mu}_k | \mathbf{y}_0^k, \mathbf{u}_0^{k-1})$$

$$p(\mathbf{y}_{k+1} | \mathbf{y}_0^k, \mathbf{u}_0^k, \mathbf{d}_0^k) = p(\mathbf{y}_{k+1} | \mathbf{y}_0^k, \mathbf{u}_0^k)$$

- Note: Both identities were mathematically derived contrary to [Peterka, Automatica, January 1981] where the only first identity was simply stated as natural condition of control.

## Unified solution – cont'd

### Active detector containing dual controller design

- Using previous identities the original BRE can be rewritten to the following form

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

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

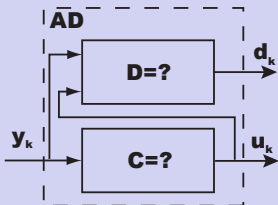
- Main new result

$$\mathbf{d}_k^* = \arg \min_{\mathbf{d}_k \in \mathcal{M}} \mathbb{E} \left\{ L_k^d(\mathbf{d}_k, \boldsymbol{\mu}_k) | \mathbf{y}_0^k, \mathbf{u}_0^{k-1}, \mathbf{d}_k \right\}$$

$$\mathbf{u}_k^* = \arg \min_{\mathbf{u}_k \in \mathcal{U}_k} \mathbb{E} \left\{ \alpha_k L_k^c(\mathbf{x}_k, \mathbf{u}_k) + V_{k+1}^* \left( \mathbf{y}_0^{k+1}, \mathbf{u}_0^k \right) | \mathbf{y}_0^k, \mathbf{u}_0^k \right\}$$

## Unified solution – cont'd

### Active detector containing dual controller design – remarks



- The optimal decision  $\mathbf{d}_k^*$  and the optimal input  $\mathbf{u}_k^*$  are chosen independently, but dual controller respect **future decisions and inputs**
- The block generating optimal decision  $\mathbf{d}_k^*$  passively utilizes given information
- The optimal input  $\mathbf{u}_k^*$  is a compromise between exciting and controlling

## Special cases

### Two special cases

- *Active detector*  
an active detector has to be designed in such a way that input signal improves fault detection only
- *Active detector containing given input signal generator*
  - a block generating input signal is given in advance
  - a block generating decision has to be designed

## The first special case

### Active detector design

- Control objective is not considered (i.e.  $\alpha_k = 0, \forall k \in \mathcal{T}$ ) and then the BRE is given as

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

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

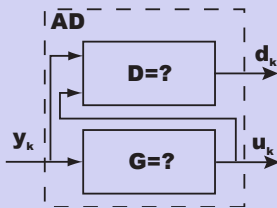
- Two minimization problems are solved simultaneously

$$\mathbf{d}_k^* = \arg \min_{\mathbf{d}_k \in \mathcal{M}} \mathbb{E} \left\{ L_k^d(\mathbf{d}_k, \boldsymbol{\mu}_k) | \mathbf{y}_0^k, \mathbf{u}_0^{k-1}, \mathbf{d}_k \right\}$$

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

## The first special case – cont'd

### Active detector – remarks

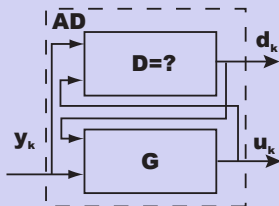


- The optimal decision  $\mathbf{d}_k^*$  and the optimal input  $\mathbf{u}_k^*$  are independent
- The optimal input  $\mathbf{u}_k^*$  only excites observed system to improve **future decisions**
- Therefore, the design of generator **G** respects future decisions



## The second special case

### Active detector containing given input signal generator



- Similarly to the first special case the control objective is not considered
- The input signal generator  $\mathbf{G}$ , **depending on decisions**, is given by the known function  $\gamma_k(\mathbf{l}_0^k, \mathbf{d}_k)$
- Note: If  $\gamma_k(\mathbf{y}_0^k, \mathbf{u}_0^{k-1})$  then a passive detection problem is solved

## The second special case – cont'd

### Active detector containing given input signal generator design

- For this special case, the BRE has the following form

$$V_k^*(\mathbf{l}_0^k) = \min_{\mathbf{d}_k \in \mathcal{M}} \left[ \mathbb{E} \left\{ L_k^d(\mathbf{d}_k, \boldsymbol{\mu}_k) + V_{k+1}^*(\mathbf{l}_0^{k+1}) \mid \mathbf{l}_0^k, \mathbf{d}_k \right\} \right]_{\mathbf{u}_k = \gamma_k(\mathbf{l}_0^k, \mathbf{d}_k)}$$

- Only one minimization problem is solved

$$\mathbf{d}_k^* = \arg \min_{\mathbf{d}_k \in \mathcal{M}} \left[ \mathbb{E} \left\{ L_k^d(\mathbf{d}_k, \boldsymbol{\mu}_k) + V_{k+1}^*(\mathbf{l}_0^{k+1}) \mid \mathbf{l}_0^k, \mathbf{d}_k \right\} \right]_{\mathbf{u}_k = \gamma_k(\mathbf{l}_0^k, \mathbf{d}_k)}$$

$$\mathbf{u}_k^* = \gamma_k(\mathbf{l}_0^k, \mathbf{d}_k^*)$$

## Numerical example

Description of the observed system 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

### Cost 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 given generator  $u_k = \gamma_k(d_k)$  for the second special case

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

## Numerical example – cont'd

### Comparison of active detectors

- Active detector containing given input generator (ADGG)
  - Active detector (ADG)
  - Active detector containing dual controller (ADC)
- 
- These approaches are compared by means of average number of wrong decisions (NWD)

### Monte Carlo simulation results

	ADGG	ADG	ADC
NWD	1.2138	1.0931	1.1008

# Conclusion

## Summary

- A problem of simultaneous active fault detection and dual control was formulated and solved as an optimization problem
- It is shown that the optimal active detector containing dual controller consists of an optimal passive detector and an optimal dual controller which excites and controls the system
- Two interesting special cases were also discussed
  - Optimal active detector without fixed input signal generator
  - Optimal active detector with fixed input signal generator