

Model-Based Fault Diagnosis - An Introduction

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Model-Based Fault Diagnosis, Isolation and Identification

What is **Fault Diagnosis**?

Identify that a system has an imperfect behaviour caused (most likely) by a fault.

What is **Fault Isolation**?

Identify location of the fault, (i.e. find out which component has become faulty.)

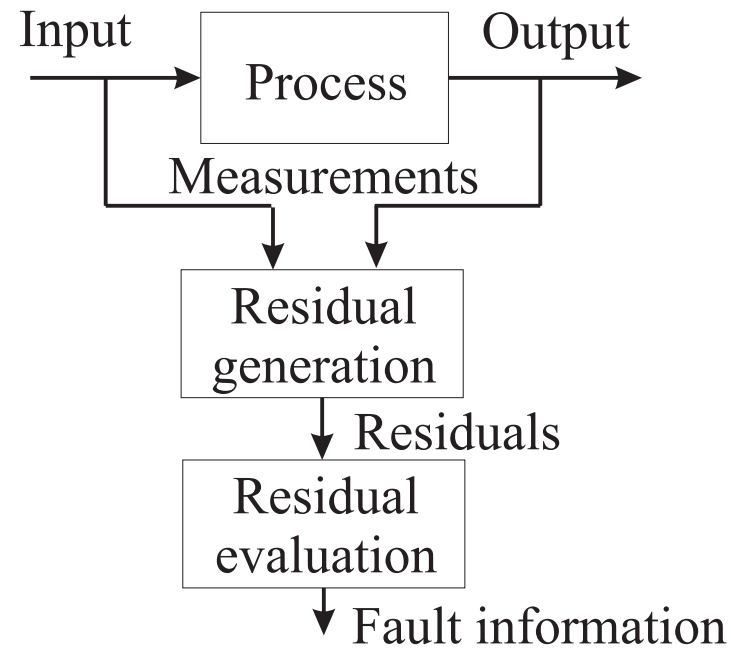
What is **Fault Identification**?

Estimate the size and type of the fault (i.e. malfunction or catastrophic breakdown)

Analytical vs Hardware redundancy for FDI

- **Hardware (or physical/parallel) redundancy.**
 - Uses multiple lanes of sensors/actuators/computer/software....
 - Requires extra hardware and a voting system to detect mismatch between redundant components.
- **Analytical (functional) redundancy.**
 - Uses redundant analytical relationships between measured variables.
 - No extra hardware is required.

Structure of model-based FDI system.



Structure of model-based FDI system.

1. Residual generation.

- This block generates residual signals using available inputs and outputs from the monitored system.
- It should be normally zero or close to zero under no fault condition, whilst is distinguishably different from zero when a fault occurs.

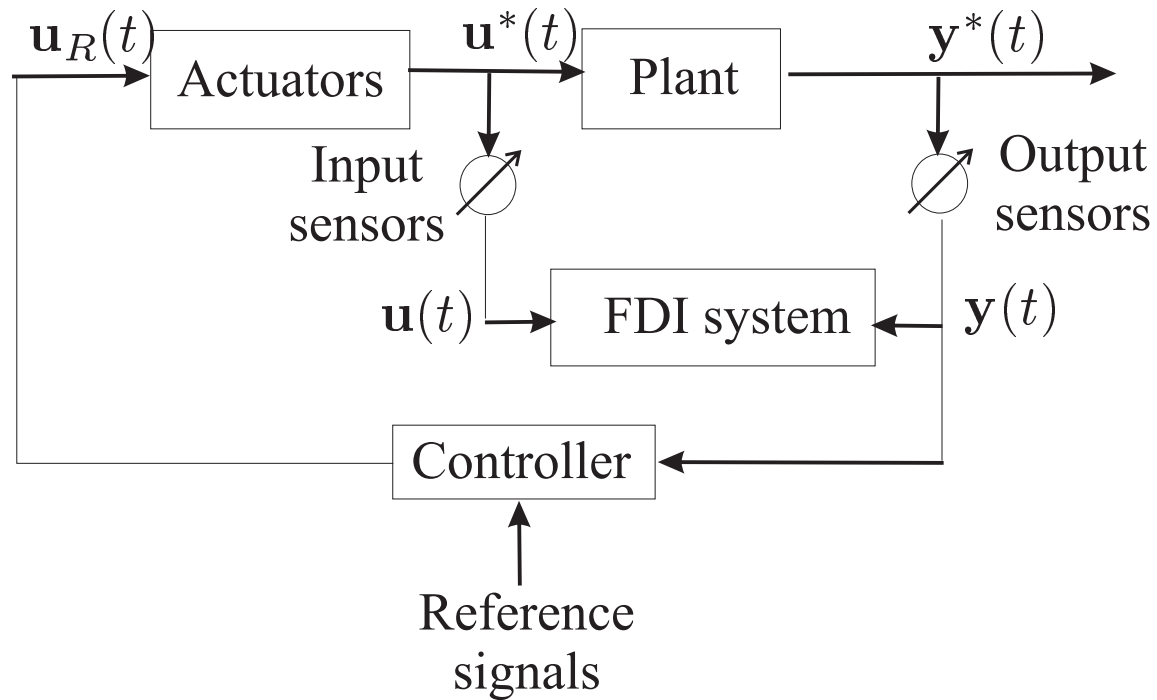
2. Residual evaluation.

- This block examines residuals for the likelihood of faults applying a decision rule.

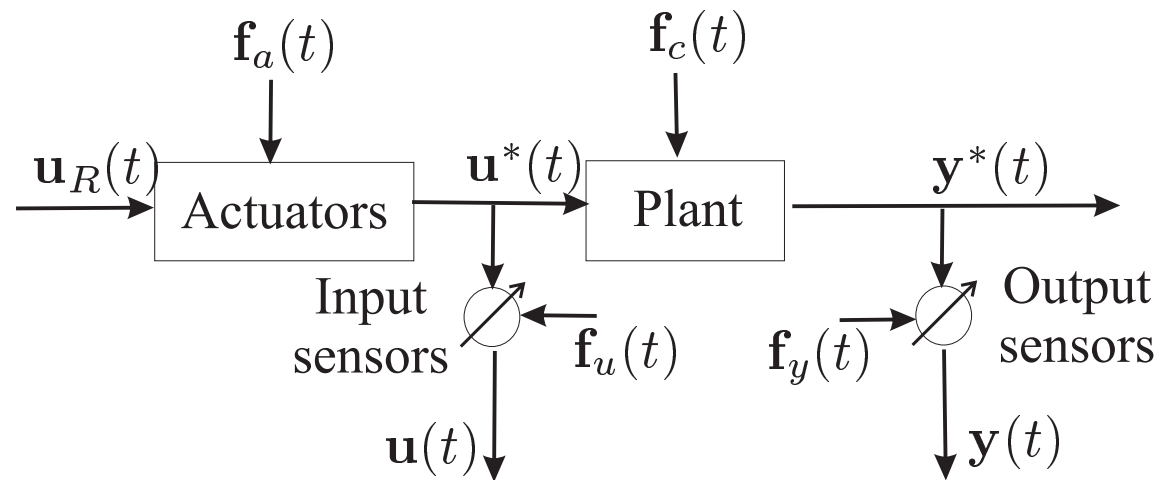
Modeling of Faulty Systems

- **Open-loop system.**
 - Input and output measurements are supposed completely available for FDI purposes.
 - We do not consider the controller behaviour in the design of a fault diagnosis scheme.
- **Closed-loop system.**
 - Input measurements are not directly available.
 - (Robust) controller may masks fault effects.
 - Research still open.

Fault diagnosis and control loop.



The monitored system and fault topology.



Linear Plant

- Consider a state-space description of a linear plant in fault free case

$$\begin{cases} \mathbf{x}(t+1) &= \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}^*(t) \\ \mathbf{y}^*(t) &= \mathbf{C}\mathbf{x}(t) \end{cases}$$

- and with faults:

$$\begin{cases} \mathbf{x}(t+1) &= \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}_R(t) + \mathbf{B}\mathbf{f}_a(t) + \mathbf{f}_c(t) \\ \mathbf{y}(t) &= \mathbf{C}\mathbf{x}(t) + \mathbf{f}_y(t) \end{cases}$$

The general case

- State-space model

$$\begin{cases} \mathbf{x}(t+1) &= \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t) + \mathbf{L}_1\mathbf{f}(t) \\ \mathbf{y}(t) &= \mathbf{C}\mathbf{x}(t) + \mathbf{L}_2\mathbf{f}(t) \end{cases}$$

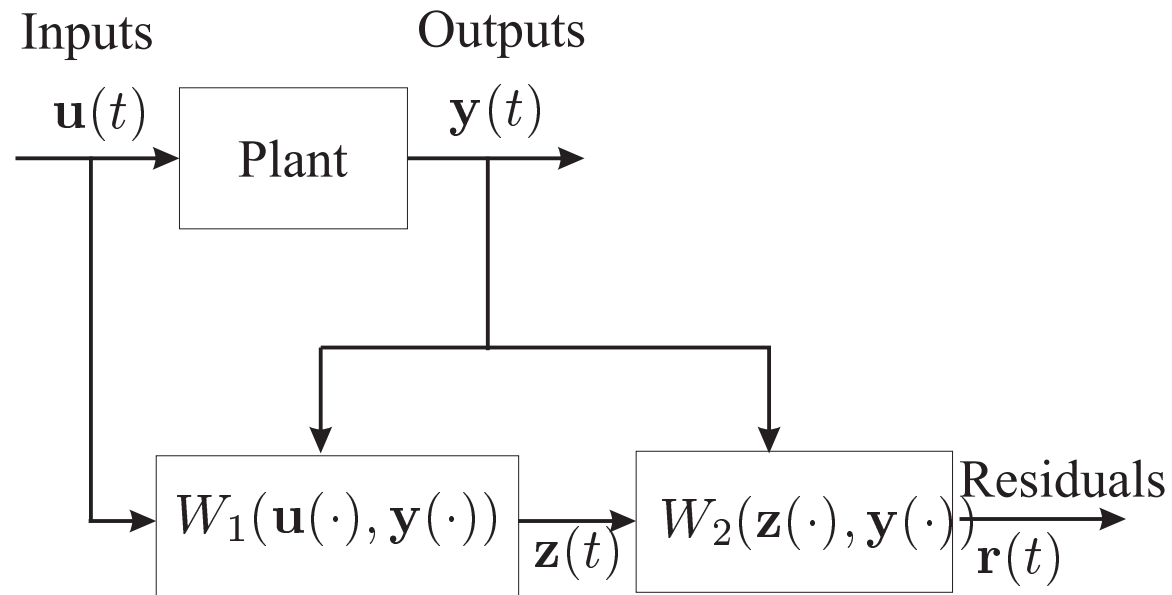
- Input-Output model

$$\mathbf{y}(s) = \mathbf{G}_{yu}(s)\mathbf{u}(s) + \mathbf{G}_{yf}(s)\mathbf{f}(s)$$

where

$$\begin{cases} \mathbf{G}_{yu}(s) &= \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{B} \\ \mathbf{G}_{yf}(s) &= \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{L}_1 + \mathbf{L}_2 \end{cases}$$

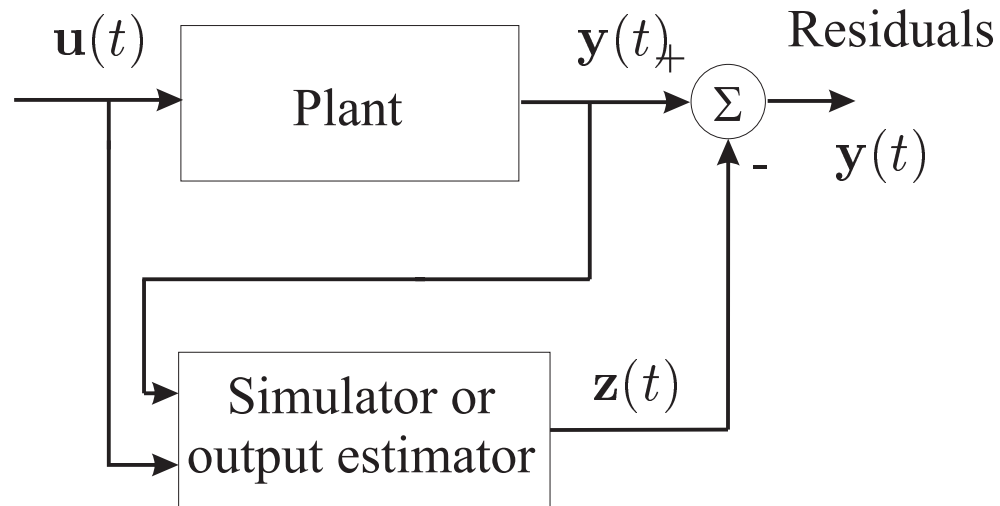
Residual Generator Structure



$$\begin{cases} \mathbf{z}(t) &= W_1(\mathbf{u}(\cdot), \mathbf{y}(\cdot)) \\ \mathbf{r}(t) &= W_2(\mathbf{z}(\cdot), \mathbf{y}(\cdot)) \end{cases}$$

Fault free: $\mathbf{r}(t) = \mathbf{0}$ Faulty: $\mathbf{r}(t) \neq \mathbf{0}$

Residual generator via output estimator.



$$W_1(\mathbf{u}(\cdot), \mathbf{y}(\cdot)) = \mathbf{M}\mathbf{y}(t)$$

$$W_2(\mathbf{z}(\cdot), \mathbf{y}(\cdot)) = \mathbf{W}(\mathbf{z}(t) - \mathbf{M}\mathbf{y}(t))$$

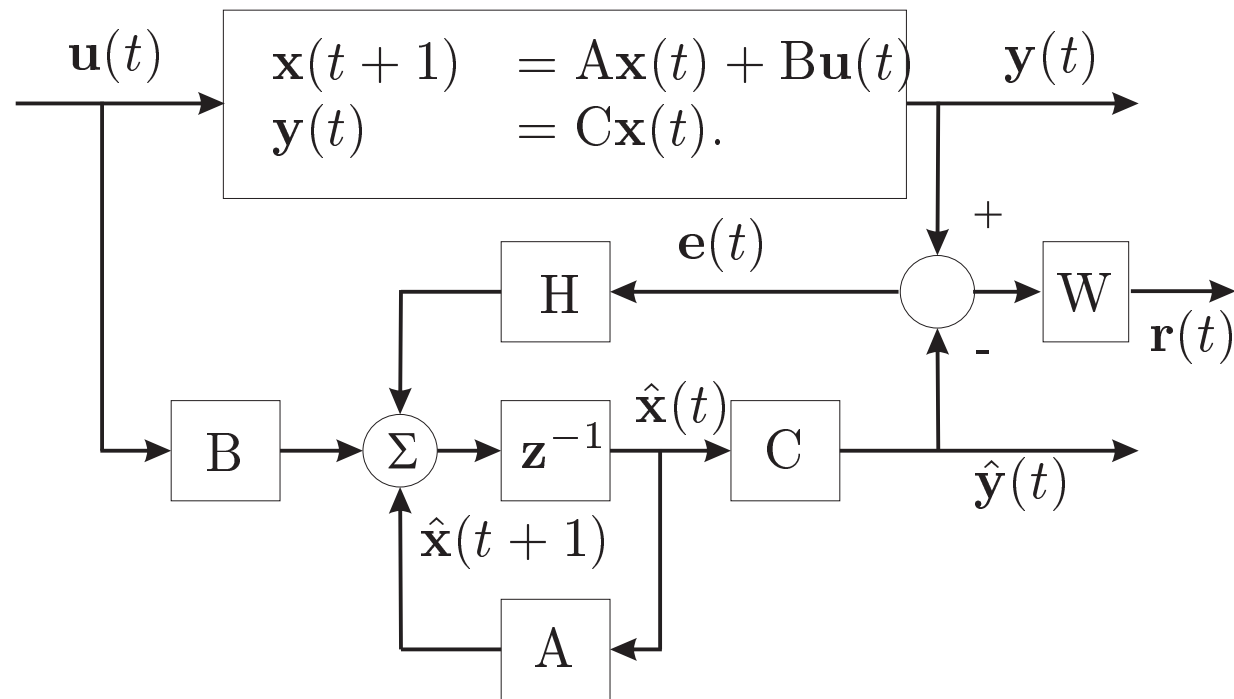
Fault detection.

- Residual limit value checking:

$$\begin{cases} \mathbf{r}(t) \leq \epsilon & \text{for } \mathbf{f}(t) = \mathbf{0} \\ \mathbf{r}(t) > \epsilon & \text{for } \mathbf{f}(t) \neq \mathbf{0} \end{cases}$$

- If the residual exceeds the threshold, a fault may be occurred.

Residual Generation with Dynamic Observers



$$\text{Observer: } \begin{cases} \hat{\mathbf{x}}(t+1) &= A\hat{\mathbf{x}}(t) + B\mathbf{u}(t) + H\mathbf{e}(t) \\ \mathbf{e}(t) &= \mathbf{y}(t) - C\hat{\mathbf{x}}(t). \end{cases}$$

State estimation error

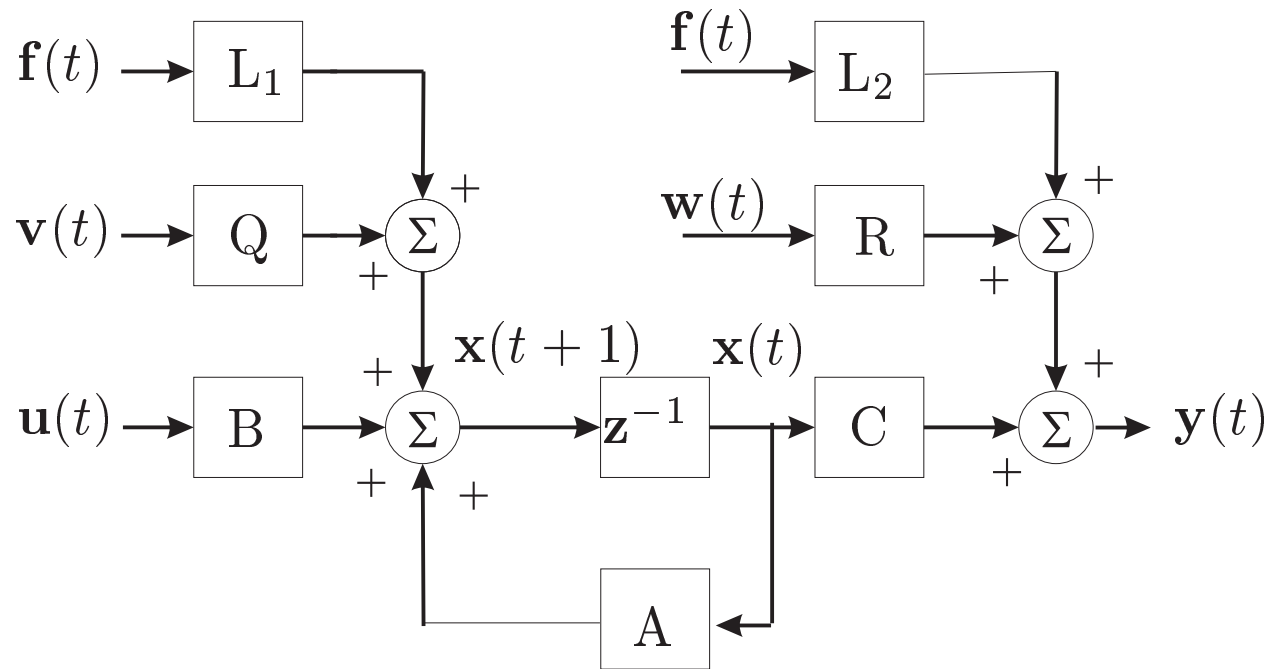
$$\begin{cases} \mathbf{e}_x(t) & = \mathbf{x}(t) - \hat{\mathbf{x}}(t) \\ \mathbf{e}_x(t+1) & = (\mathbf{A} - \mathbf{H}\mathbf{C})\mathbf{e}_x(t). \end{cases}$$

The state error $\mathbf{e}_x(t)$ vanishes asymptotically

$$\lim_{t \rightarrow \infty} \mathbf{e}_x(t) = \mathbf{0}$$

if the observer is stable (by means of proper design of the observer feedback \mathbf{H}).

Faults in the observer scheme



$$\begin{cases} \mathbf{x}(t+1) &= \mathbf{A}\mathbf{x}(t) + \mathbf{B}\mathbf{u}(t) + \mathbf{Q}\mathbf{v}(t) + \mathbf{L}_1\mathbf{f}(t) \\ \mathbf{y}(t) &= \mathbf{C}\mathbf{x}(t) + \mathbf{R}\mathbf{w}(t) + \mathbf{L}_2\mathbf{f}(t) \end{cases}$$

State estimation error in presence of faults - I

- Hypothesis: $\mathbf{v}(t) = \mathbf{0}$ and $\mathbf{w}(t) = \mathbf{0}$

$$\begin{aligned}\mathbf{e}_x(t+1) &= (\mathbf{A} - \mathbf{H}\mathbf{C})\mathbf{e}_x(t) + \mathbf{L}_1\mathbf{f}(t) - \mathbf{H}\mathbf{L}_2\mathbf{f}(t) \\ \mathbf{e}(t) &= \mathbf{C}\mathbf{e}_x(t) + \mathbf{L}_2\mathbf{f}(t)\end{aligned}$$

- where
 - $\mathbf{C}\mathbf{e}_x(t)$ represent faults modulated by system dynamics.
 - $\mathbf{L}_2\mathbf{f}(t)$ represents fault affecting output directly.
- In fact, considering the Z - *trans.* of $\mathbf{e}(t)$:

$$\mathbf{e}(z) = \mathbf{C}(z\mathbf{I} - \mathbf{A} + \mathbf{H}\mathbf{C})^{-1}(\mathbf{L}_1\mathbf{f}(z) - \mathbf{H}\mathbf{L}_2\mathbf{f}(z)) + \mathbf{L}_2\mathbf{f}(z)$$

State estimation error in presence of faults - II

- If faults appear as changes ΔA or ΔB :

$$\begin{cases} \mathbf{x}(t+1) &= (\mathbf{A} + \Delta\mathbf{A})\mathbf{x}(t) + (\mathbf{B} + \Delta\mathbf{B})\mathbf{u}(t) \\ \mathbf{y}(t) &= \mathbf{C}\mathbf{x}(t) \end{cases}$$

while the state $\mathbf{e}_x(t)$ and the output estimation $\mathbf{e}(t)$ errors

$$\begin{cases} \mathbf{e}_x(t+1) &= (\mathbf{A} - \mathbf{H}\mathbf{C})\mathbf{e}_x(t) + \Delta\mathbf{A}\mathbf{x}(t) + \Delta\mathbf{B}\mathbf{u}(t) \\ \mathbf{e}(t) &= \mathbf{C}\mathbf{e}_x(t). \end{cases}$$

- The changes ΔA and ΔB are then **multiplicative faults**.

Parity relation (vector) method

- The FDI problem can be posed either in **Transfer Function** and **State Space** forms.
- Let's consider the **Transfer Function approach**

$$y(z) = M(z)u(z) + S(z)p(z)$$

$y(t)$ system outputs
 $u(t)$ known (observed) inputs
 $p(t)$ (additive) unknown inputs

$u(z)$ and $p(z)$ can be further decomposed into

$$u(z) \begin{cases} u_c(z) & \text{Controlled input} \\ u_m(z) & \text{measured inputs} \end{cases}$$

$$p(z) \begin{cases} p_f(z) & \text{additive faults (input and output sensor faults, actuator faults)} \\ p_d(z) & \text{additive disturbance} \end{cases}$$

The system becomes:

$$y(z) = [M_c(z) \ M_m(z)] \begin{bmatrix} u_c(z) \\ u_m(z) \end{bmatrix} + [S_f(z) \ S_d(z)] \begin{bmatrix} p_f(z) \\ p_d(z) \end{bmatrix}$$

Parametric (system) faults and model error can be represented by

$$y(z) = M(z)u(z) + \Delta M(z)u(z)$$

$$\Delta M(z) \begin{cases} \Delta M_f(z) & \text{parametric faults} \\ \Delta M_d(z) & \text{modelling errors} \end{cases}$$

leading to:

$$y(z) = M(z)u(z) + [\Delta M_f(z) \Delta M_d(z)]u(z)$$

Residual generation for additive faults

- Residual generation:

$$r(z) = W(z)[y(t) - M(z)u(z)]$$

where $r(z)$ is the residual and $W(z)$ is the residual generation matrix.

- substituting $y(t)$ yields the internal form

$$r(z) = W(z)S(z)p(z)$$

- One may specify the desired response as

$$r(z) = Z(z)p(z)$$

- comparing this to $r(z) = W(z)S(z)p(z)$, we obtain the **implementation** equation

$$W(z)S(z) = Z(z)$$

where $S(z)$ and $Z(z)$ are given and $W(z)$ is to be found.

The equivalence of linear residual generators.

- Any residual generator can be described as

$$r(z) = W(z)y(z) + V(z)u(z)$$

- $r(z)$ should be zero in absence of faults or disturbance. This requires

$$W(z)M(z) + V(z) = 0$$

- leading to the “parity relation”:

$$r(z) = W(z)[y(t) - M(z)u(z)]$$

Robust Residual Generation Problem

- Mathematical description of the monitored system that includes any kind of modelling uncertainty:

$$\begin{aligned}\dot{x} &= (A + \Delta A)x(t) + (B + \Delta B)u(t) + E_1d(t) + L_1f(t) \\ y(t) &= (C + \Delta C)x(t) + E_2d(t) + L_2f(t)\end{aligned}$$

- Transfer function:

$$y(s) = (G_u(s) + \Delta G_u(s))u(s) + G_d(s)d(s) + G_f(s)f(s)$$

- residual generator:

$$r(s) = H_y(s)G_f(s)f(s) + \overbrace{H_y(s)\Delta G_u(s) + H_y(s)G_d(s)d(s)}^{\text{Spurious terms}}$$

Robustness to disturbances

- Total decoupling to disturbance:

$$H_y(s)G_d(s) = 0$$

- If only a partial decoupling can be achieved, then a **performance index** can be defined:

$$J = \frac{H_y(j\omega)G_d(j\omega)}{H_y(j\omega)G_f(j\omega)}$$

Adaptive threshold in robust FDI

- Residual uncertainty arising from modelling errors:

$$r(s) = H_y(s)\Delta G_u(s)$$

- Assuming that the modelling errors are bounded:

$$\|\Delta G_u(s)\| \leq \delta$$

- fault free residual will be bounded as

$$\begin{aligned} \|r(j\omega)\| &= \|H_y(j\omega)\Delta G_u(j\omega)u(j\omega)\| \leq \\ &\leq \|H_y(j\omega)u(j\omega)\| \|\Delta G_u(j\omega)\| \leq \delta \|H_y(j\omega)u(j\omega)\| \end{aligned}$$

- Therefore an adaptive threshold can be generated as

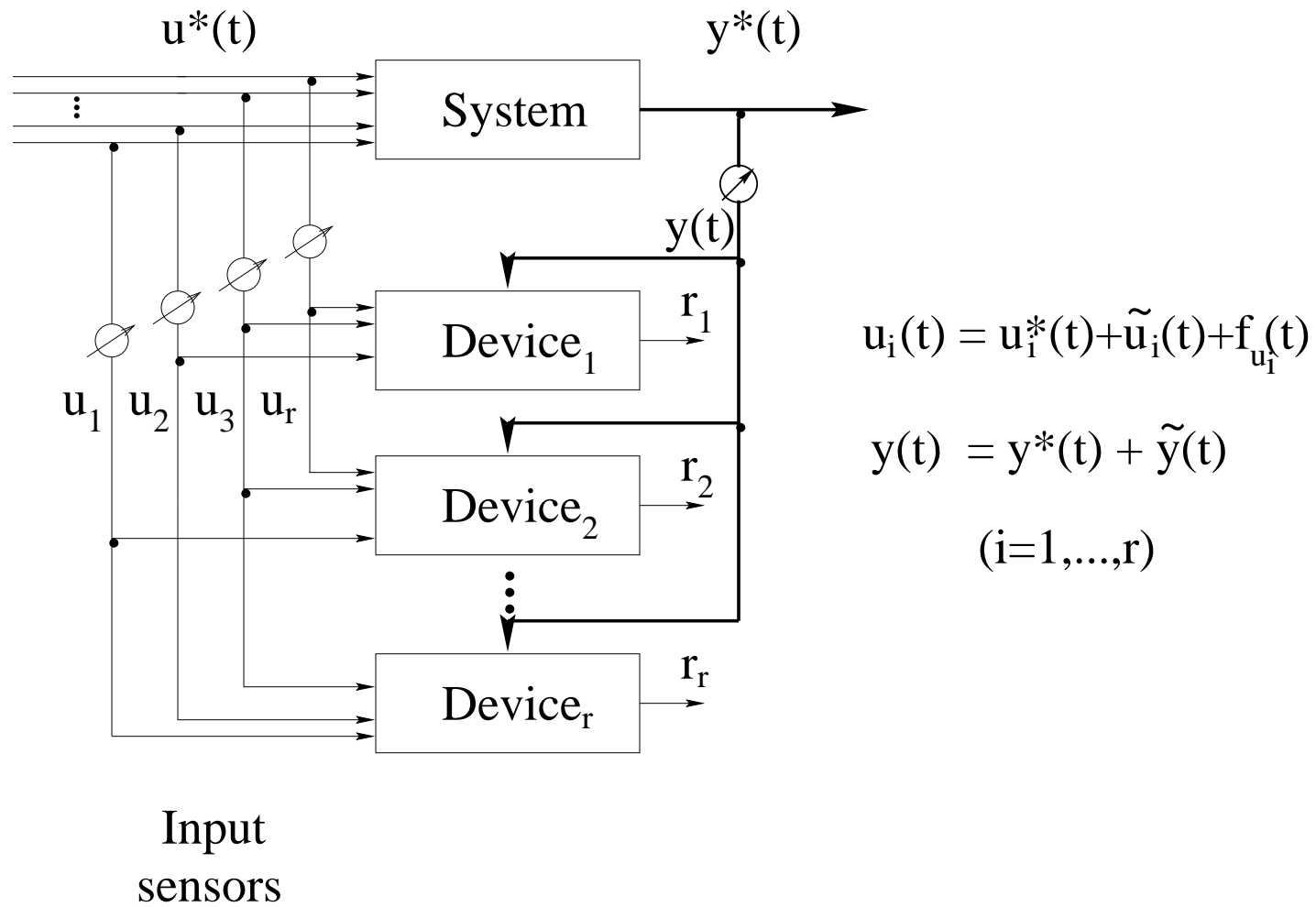
$$T(s) = \delta H_y(s)u(s)$$

- a fault is declared if:

$$\|r(t)\| \leq \|T(t)\|$$

Unknown Input Observer

- The i -th device is driven by all but the i -th input sensor and all outputs of the system.
- The residual function which is sensitive to all but the i -th input sensor fault.
- A fault on the i -th input sensor affects all the residual functions except that of the device which is insensitive to the i -th input.



Output sensor fault diagnosis

