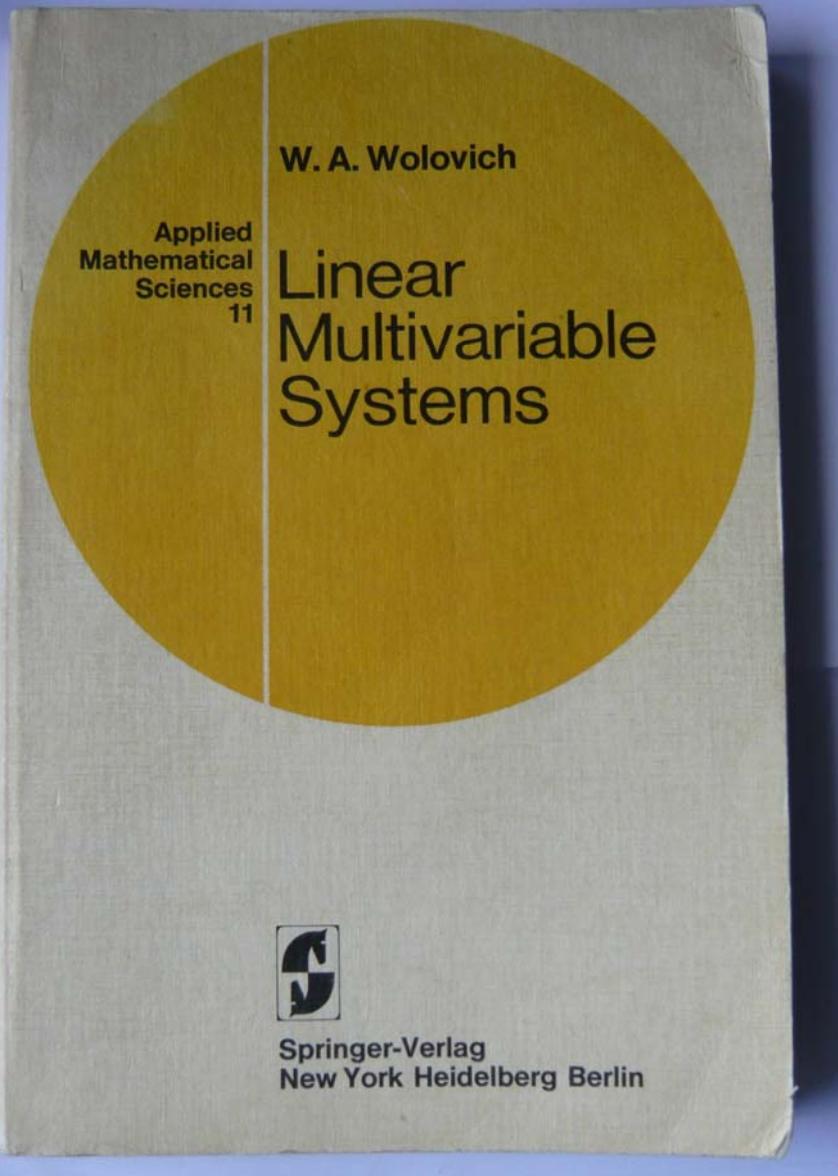




# Sensor classification for control and diagnosis problems: a structural approach

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# In Honor of **Bill Wolovich**

*My old book*

*When I was PhD student  
in 1974*

# **In Honor of Bill Wolovich**

- His results on polynomial factorizations was a starting point for me to work on system theory (together with results of V. Kucera)
- His results on the prior knowledge for adaptive control (interactor) was also a starting point for me to work on stability analysis of adaptive systems (together with results of S. Morse)

# ***Introduction***

## *Aim of this talk*

- Sensor network design problem for observability and diagnosis
- Structural modeling of dynamical systems
- Sensor classification (useless, essential)
- Quantification of the criticity of useful sensors
- Graph approach with low complexity
- Illustration on an example

# References

- FDI: Frank 96, Chen and Patton 99,
- Structural models: Lin 74, Murota 87, van der Woude 00, Dion Commault van der Woude 03
- Observability or FDI in this context: Lin 74, Boukhobza 06, Staroswieki 06, Commault Dion and Trinh 08

# Problem formulation

Dynamical system  $\Sigma$

$$\begin{cases} \dot{x}(t) = f(x(t), u(t)) \\ y(t) = h(x(t)) \end{cases}$$
$$u \in R^m, x \in R^n, y \in R^p$$

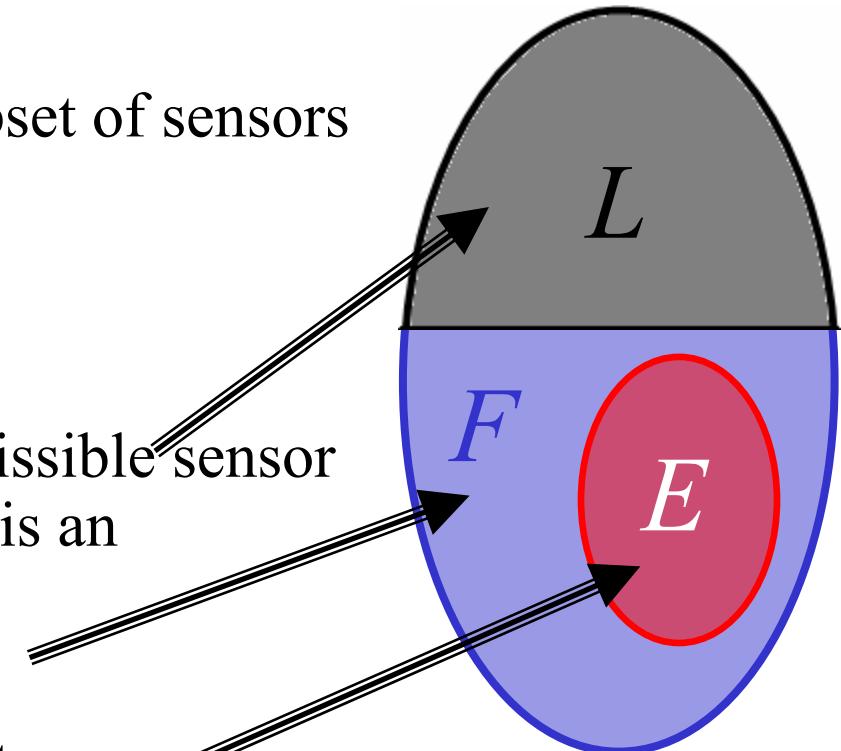
Consider a property  $P$  which is TRUE with  $\Sigma$   
(for example observability)

Problem:

Is  $P$  preserved under sensor failure?

# Sensor classification

- Property  $P$ 
  - Admissible Sensor Set: Subset of sensors  $V$  such that  $P$  is true for  $V$
- Sensor classification:
  - $y^*$  is useless if for any admissible sensor set containing  $y^*$ ,  $V \setminus \{y^*\}$  is an admissible sensor set
  - Useful sensors: not useless
  - $y^*$  is essential if  $y^*$  belongs to any admissible sensor set



# **Sensor classification**

- In this talk,  $P$  will be:
  - Solvability of the Fault Detection and Isolation (FDI) Problem

# **Structured systems**

Linear systems in which the entries of the matrices in a state space representation are:

- *zeros*
- *independent parameters*

Generic properties:

valid for almost any value of the parameters

# **Associated graph**

$$\Sigma : \begin{cases} \dot{x}(t) = A_\lambda x(t) + B_\lambda u(t) \\ y(t) = C_\lambda x(t) + D_\lambda u(t) \end{cases}$$

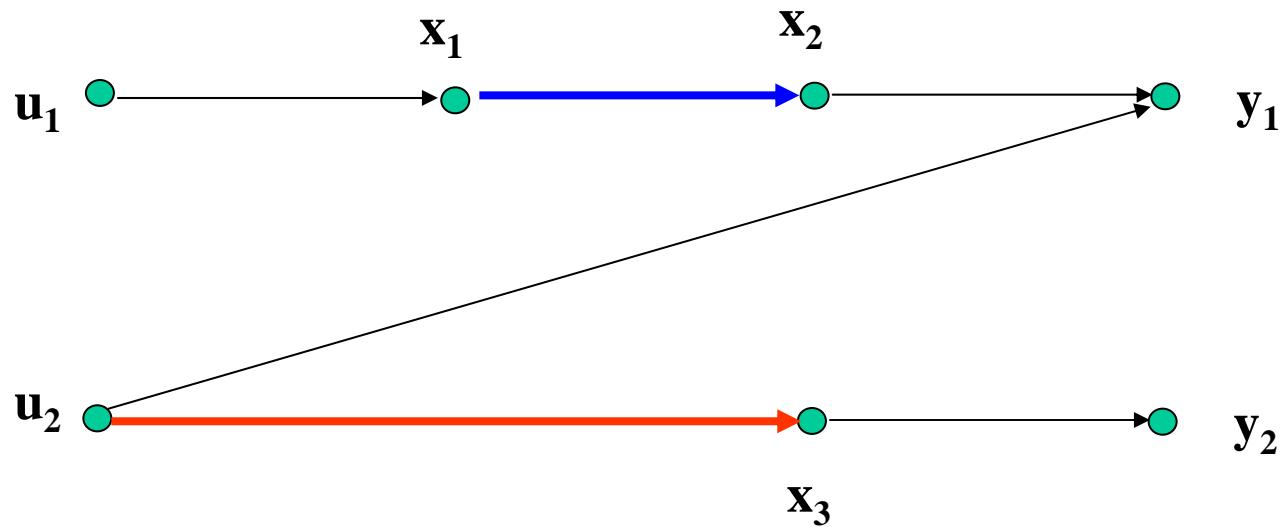
Associated graph:

- vertex set : input, state, output vertices
- edge set : corresponds to non zero entries in matrices  
(as many edges as parameters  $\lambda_i$ )

# Associated graph

Example

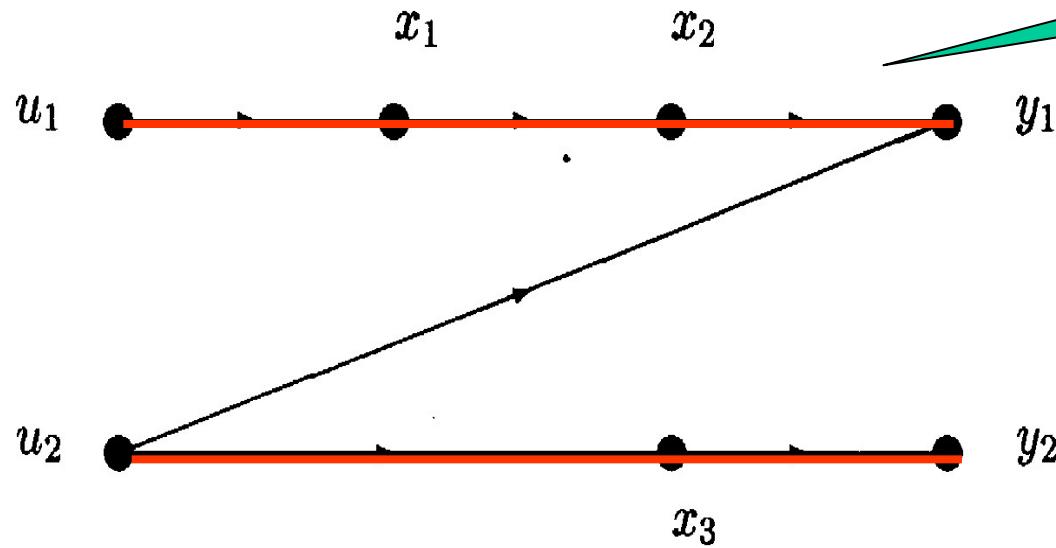
$$A_\lambda = \begin{pmatrix} 0 & 0 & 0 \\ \lambda_1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \quad B_\lambda = \begin{pmatrix} \lambda_2 & 0 \\ 0 & 0 \\ 0 & \lambda_3 \end{pmatrix}, \quad C_\lambda = \begin{pmatrix} 0 & \lambda_4 & 0 \\ 0 & 0 & \lambda_5 \end{pmatrix}, \quad D_\lambda = \begin{pmatrix} 0 & \lambda_6 \\ 0 & 0 \end{pmatrix}.$$



# Generic rank of a transfer matrix

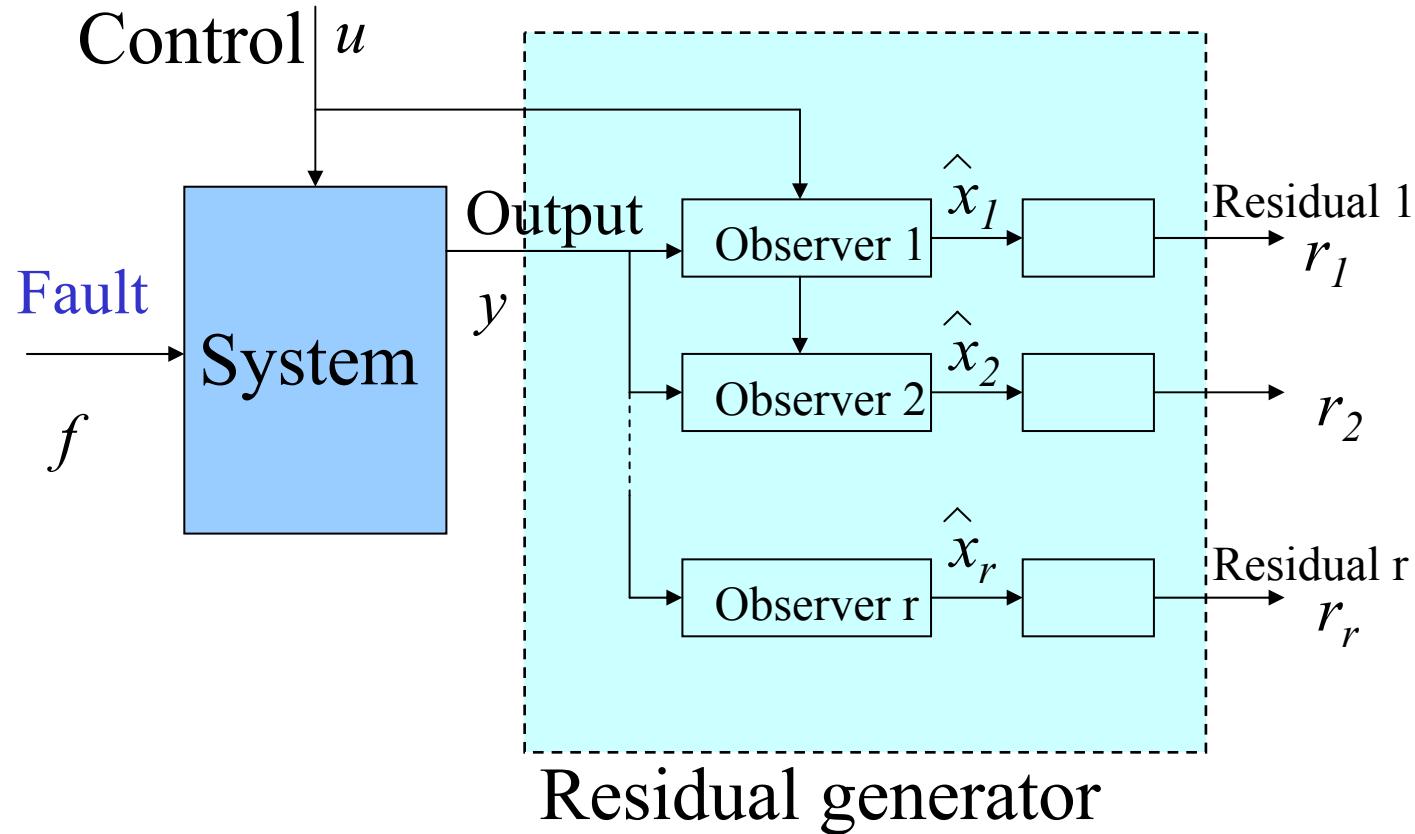
- Generic rank of  $T(s)$  = Maximal number of vertex disjoint input-output paths (*van der Woude 90*)
- (Maximal linking)

Generic rank = 2



$$T_\lambda(s) = \begin{pmatrix} \frac{\lambda_1 \lambda_2 \lambda_4}{s^2} & \lambda_6 \\ 0 & \frac{\lambda_3 \lambda_5}{s} \end{pmatrix}$$

# Observer-based FDI problem



# **Observer-based FDI problem**

Linear System

$$\Sigma : \begin{cases} \dot{x}(t) = Ax(t) + Lf(t) + Bu(t) \\ y(t) = Cx(t) + Mf(t) \end{cases}$$

Observer  $i$

$$\dot{\hat{x}}^i(t) = A\hat{x}^i(t) + K^i(y(t) - C\hat{x}^i(t)) + Bu(t)$$

Residual  $i$

$$r_i(t) = Q^i(y(t) - C\hat{x}^i(t))$$

# **FDI Problem**

With the available sensors, make the system **observable** and

$$r(s) = \begin{bmatrix} t_{11}(s) & 0 & \cdots & 0 \\ 0 & t_{22}(s) & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & t_{rr}(s) \end{bmatrix} [f(s)]$$

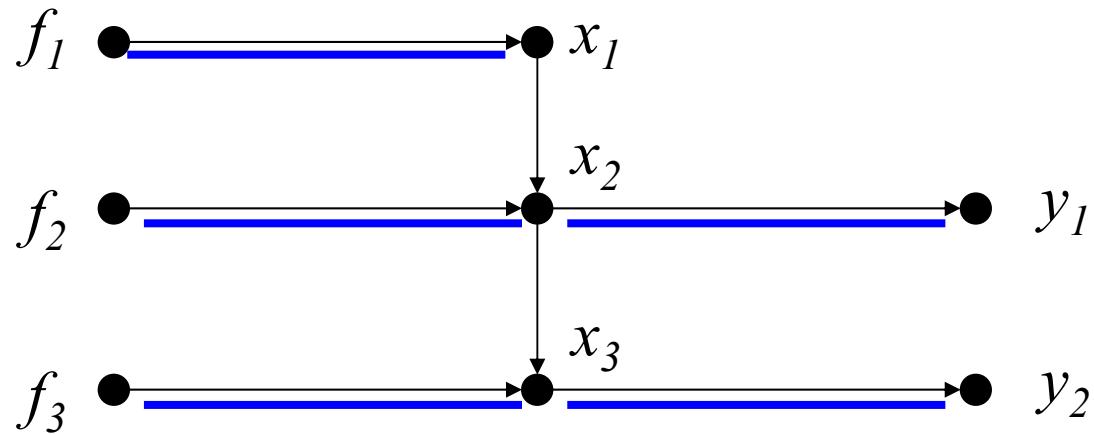
FDI problem structurally solvable iff (*Commault et al 02*):

1. The system is **structurally observable**
2.  $k = r$  (**rank condition**)

$r$  : number of faults

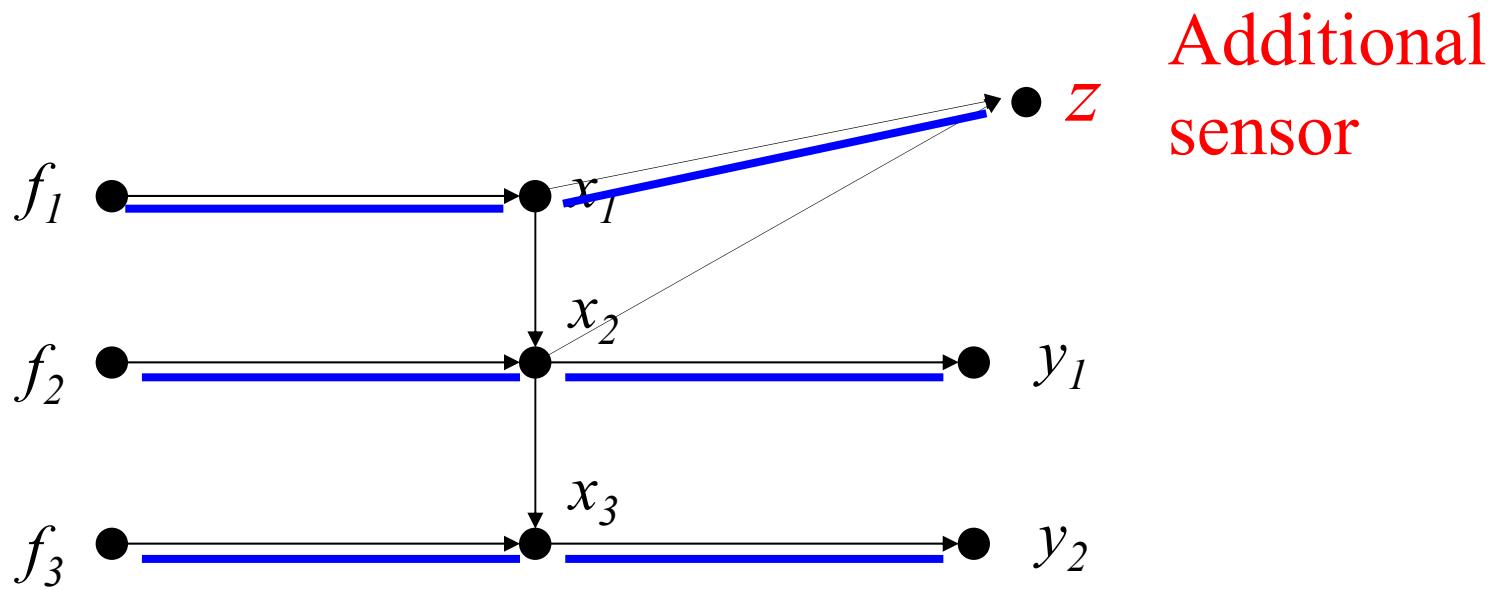
$k$  : maximal number of vertex disjoint fault-output paths

# Example



Observable but  $k=2$ ,  $r=3$   
Problem not solvable

# Example



Observable and  $k=3$ ,  $r=3$   
problem solvable

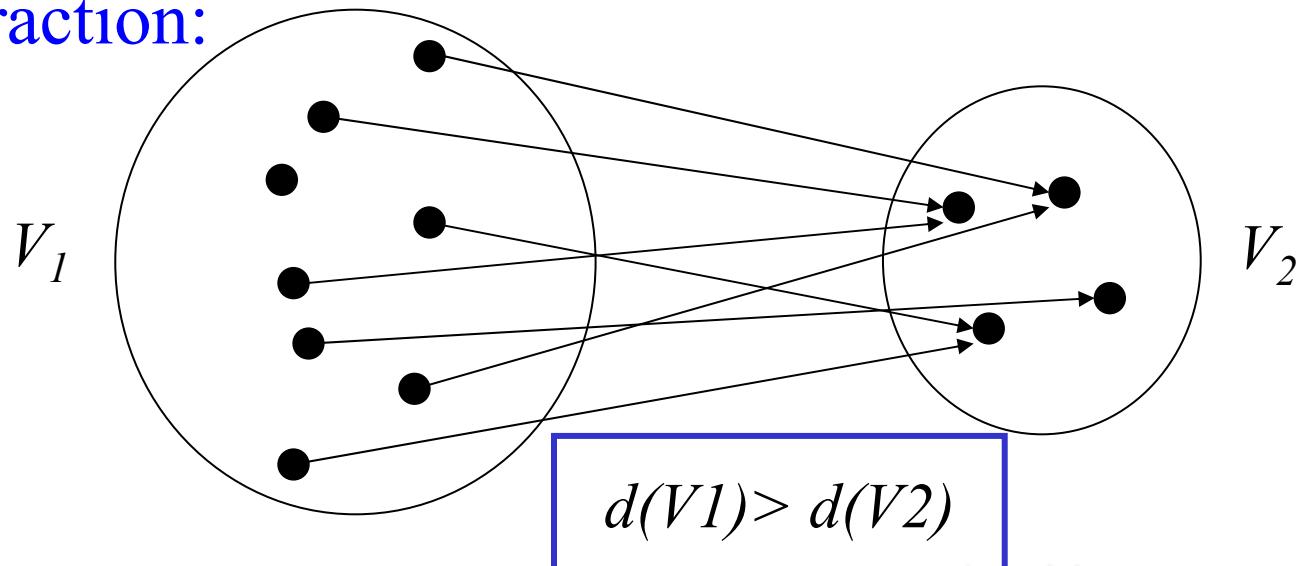
# The observability conditions

Theorem (Lin, 74)

The system is generically observable if and only if:

- There is a path from any state vertex to the outputs **(output connection)**
- There is **no contraction** in the graph

Contraction:



# **Output connection analysis**

Irreducible separators:

*Minimal sets of output vertices which when removed, disconnect some states from outputs*

# **Sensor classification for the output connection condition**

**Theorem:**

Useless sensors:

$y_i$  is useless  $\Leftrightarrow$  does not belong to an irreducible separator

Essential sensors:

$y_i$  is essential  $\Leftrightarrow$  belongs to an irreducible separator of dimension one

# Example

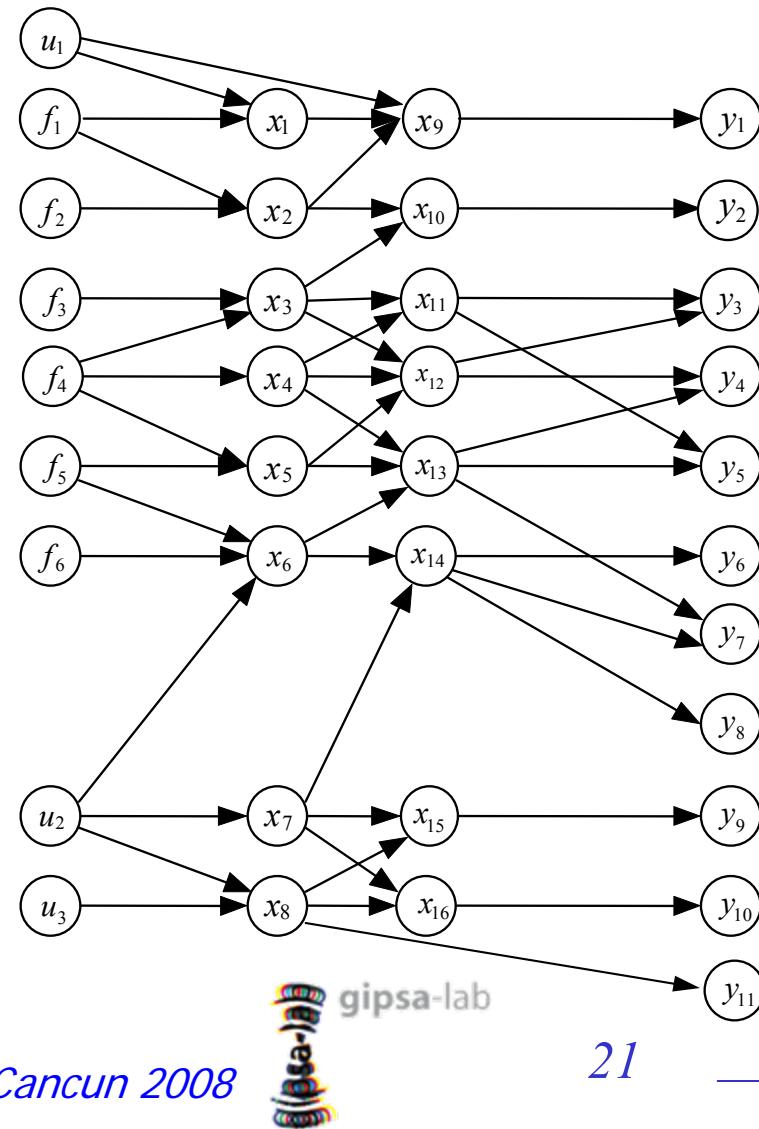
System with:

*3 control inputs*

*6 faults*

*10 sensors*

*14 states*



# For output connection

Associated graph:

Irreducible separators:

$$S_1 = \{y_1\}$$

$$S_2 = \{y_2\}$$

$$S_7 = \{y_9\}$$

$$S_8 = \{y_{10}\}$$

Essential  
sensors

$$\{y_1, y_2, y_9, y_{10}\}$$

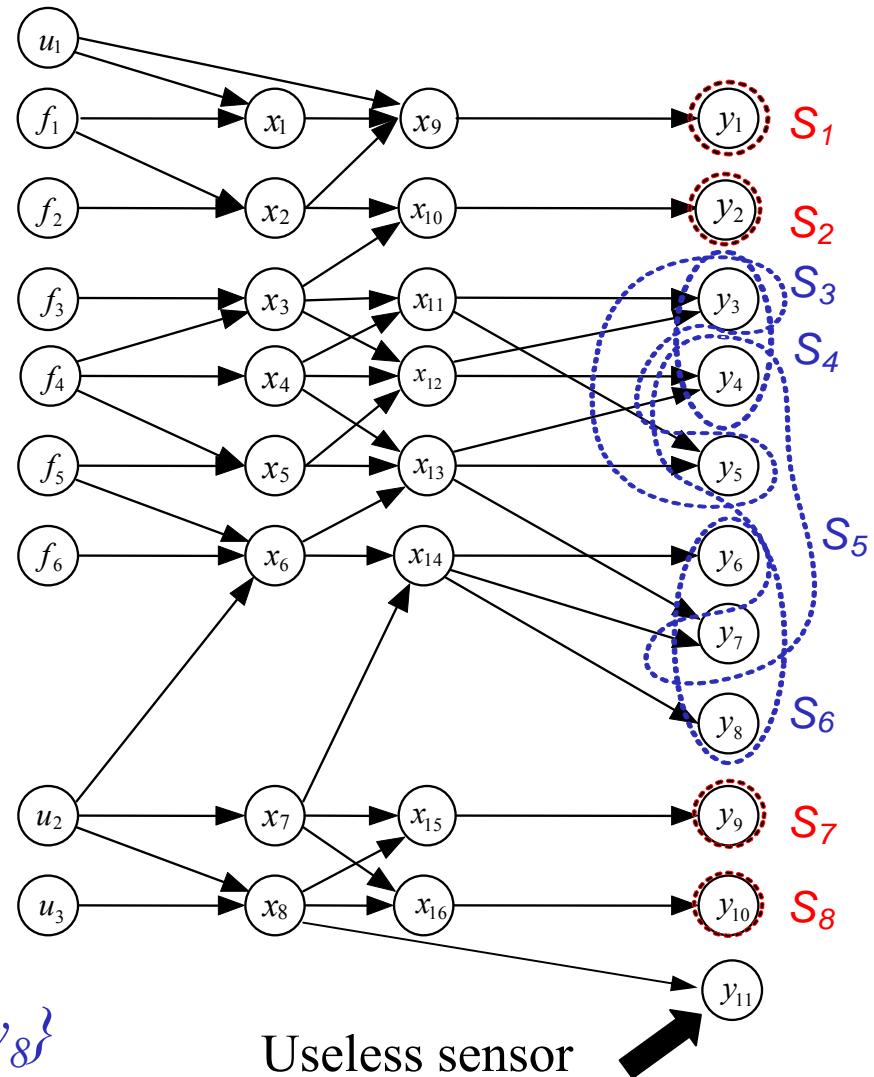
$$S_3 = \{y_3, y_5\}$$

$$S_4 = \{y_3, y_4\}$$

$$S_5 = \{y_4, y_5, y_7\}$$

$$S_6 = \{y_6, y_7, y_8\}$$

Useful  
sensors  
 $\{y_1, y_2, y_9, y_{10},$   
 $y_3, y_4, y_5, y_6, y_7, y_8\}$



# **Contraction analysis**

On the DM decomposition of the bipartite graph associated with the system

We look for matchings of maximum cardinality

No contraction when the maximal matching covers all the state vertices

# **Sensor classification for the contraction avoidance for FDI**

**Theorem:**

**Useless sensor:**

$y_i$  is useless  $\Leftrightarrow y_i$  of no use to build a maximal matching in the bipartite graph

**Essential sensor:**

$y_i$  is essential  $\Leftrightarrow$  belongs to the  $B_i$  components of the DM decomposition of the bipartite graph

# Example

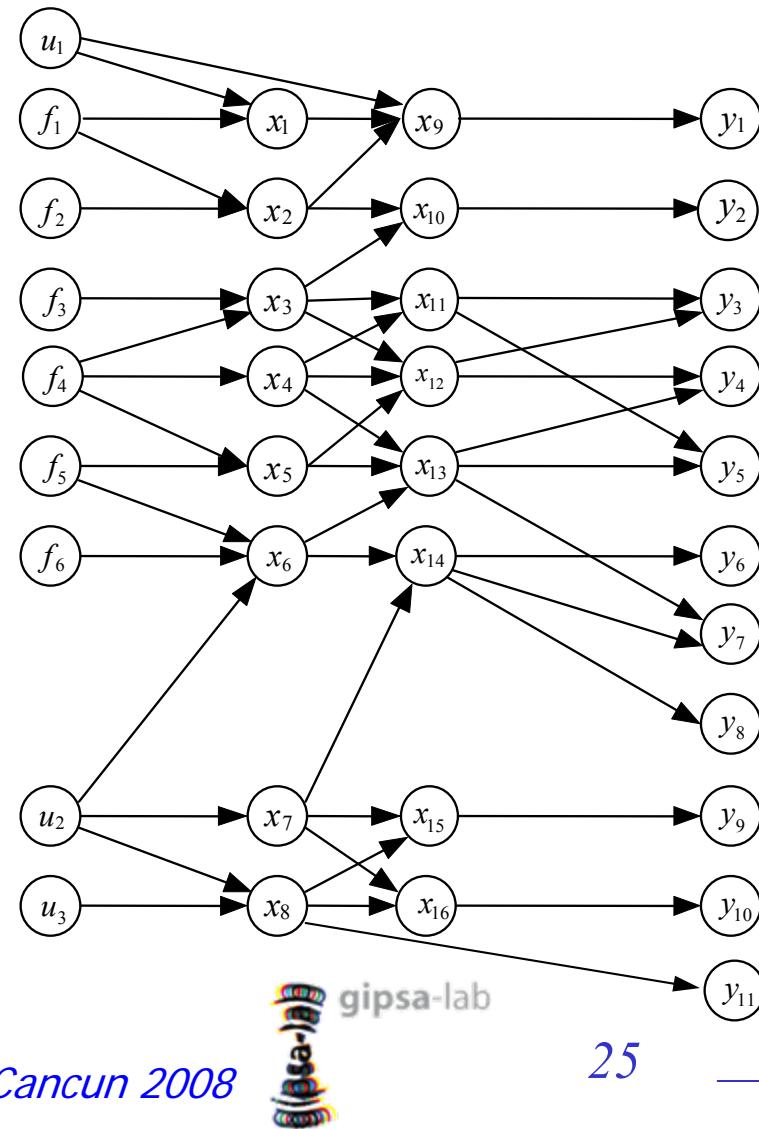
System with:

*3 control inputs*

*6 faults*

*10 sensors*

*14 states*



# For contraction avoidance

Essential sensors

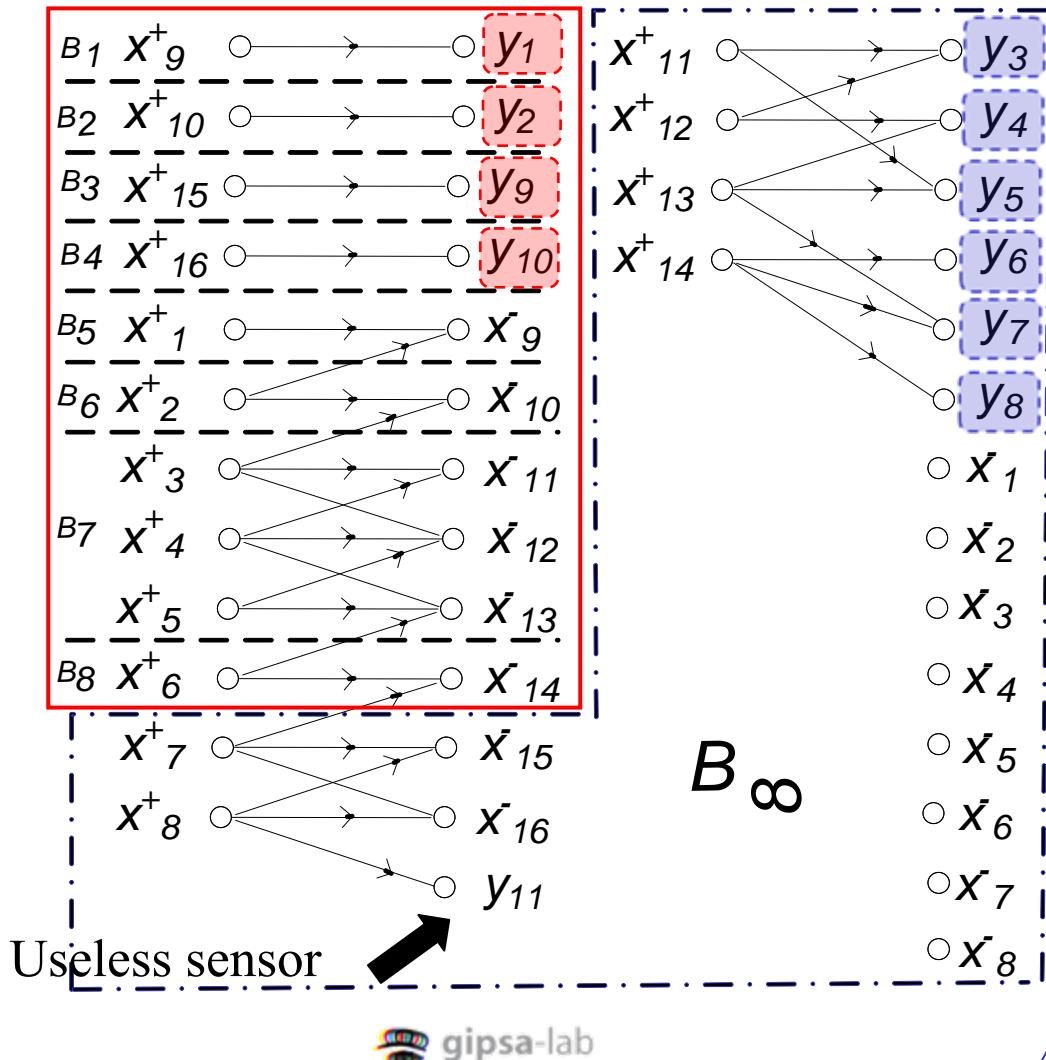
$$\{y_1, y_2, y_9, y_{10}\}$$

Useful sensors

$$\{y_1, y_2, y_9, y_{10}, y_3, y_4, y_5, y_6, y_7, y_8\}$$

Useless sensor

$$\{y_{11}\}$$



# The FDI rank condition

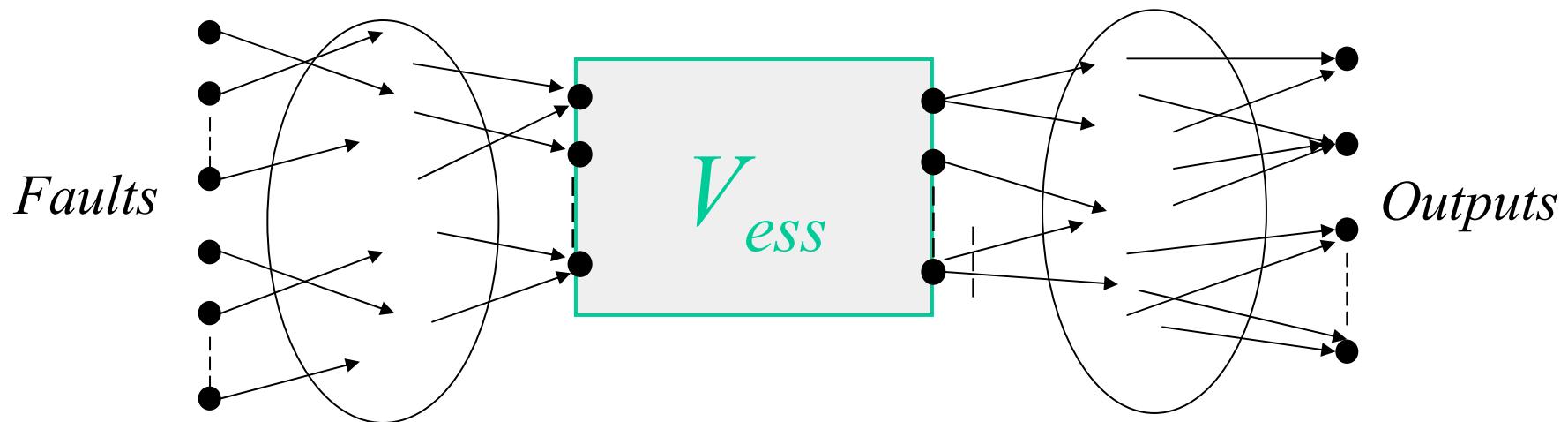
Linear observable system  $\Sigma$  with  $r$  faults and sensor set  $Y$ :

$$\Sigma : \begin{cases} \dot{x}(t) = Ax(t) + Lf(t) + Bu(t) \\ y(t) = Cx(t) + Mf(t) \end{cases}$$

Subset  $V \subseteq Y$  admissible sensor set for the rank condition of the FDI problem  
 $\Leftrightarrow$  There exists an F-V linking of size  $r$  in  $G(\Sigma_\Lambda)$  (*rank condition for FDI*)

# **Essential vertices**

Belong to any maximal size fault-output linking (F-Y linking)



# **Sensor classification for the rank condition for FDI**

**Theorem:**

Useless sensor:

$y_i$  is useless  $\Leftrightarrow$  There is no  $F$ - $y_i$  path

Essential sensor:

$y_i$  is essential  $\Leftrightarrow y_i$  belongs to  $V_{ess}$

# Example

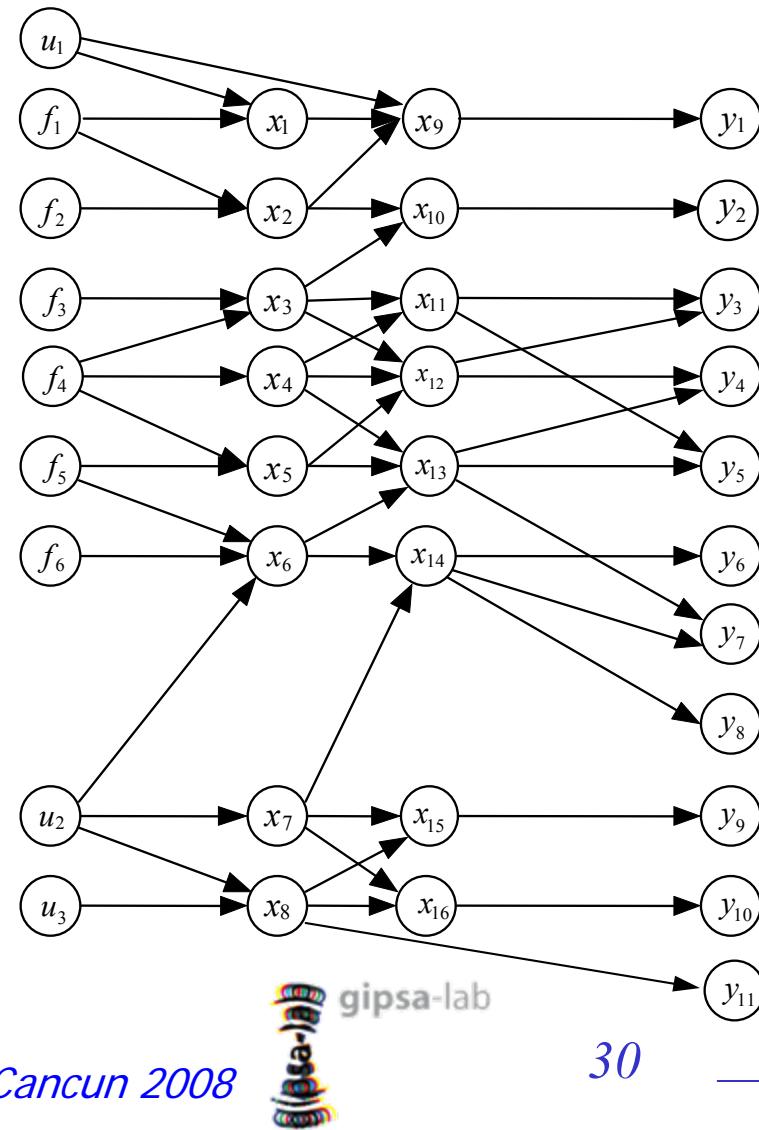
System with:

*3 control inputs*

*6 faults*

*10 sensors*

*14 states*



# FDI rank condition

System with:

*3 control inputs*

*6 faults*

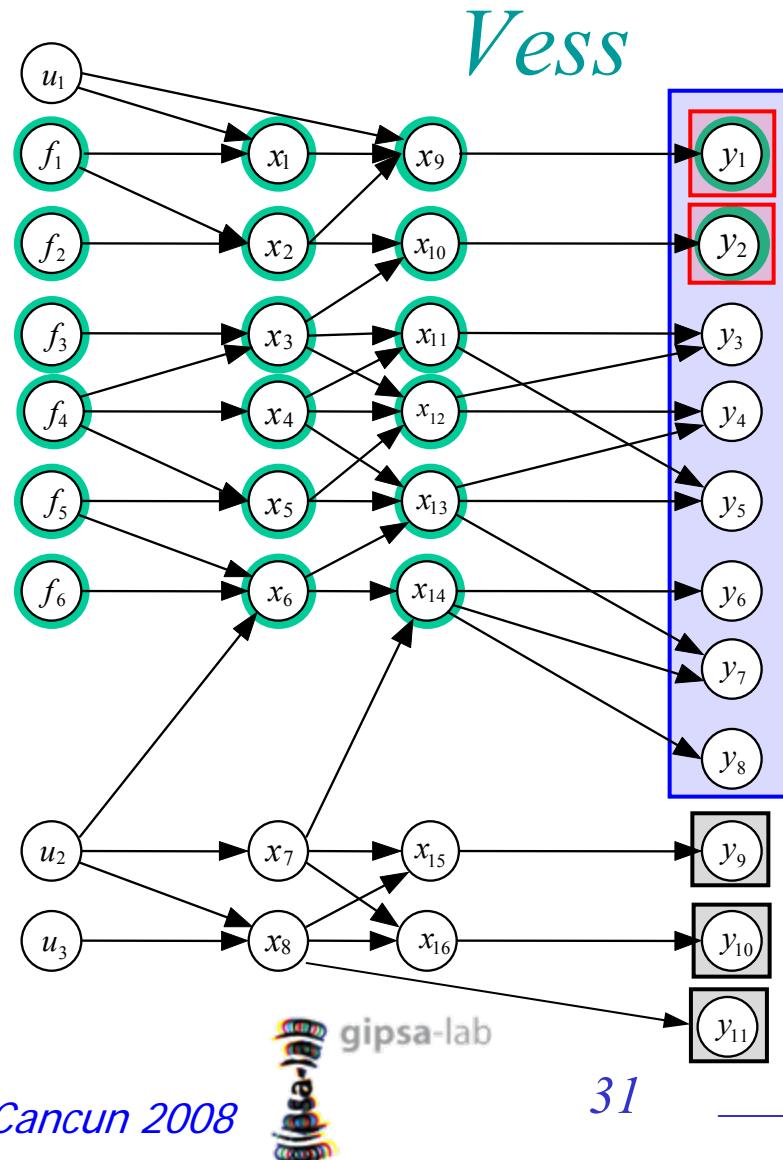
*10 sensors*

*14 states*

Essential sensors:  $y_1, y_2$

Useless sensors:  $y_9, y_{10}, y_{11}$

Useful sensors:  $y_1, \dots, y_8$



# **Complexity**

For the three problems one gets polynomial time bounded algorithms either for the determination of the set of essential sensors or for the set of useless sensors.

By standard max flow algorithms and by labeling procedures

→ The classification of sensors is polynomial

# **Index of criticity for the sensors**

$K$ = cardinality of the set of admissible sensor sets containing no useless sensors

$K_i$ = cardinality of the set of admissible sensor sets of  $K$  containing  $y_i$

*Criticality degree of  $y_i$*

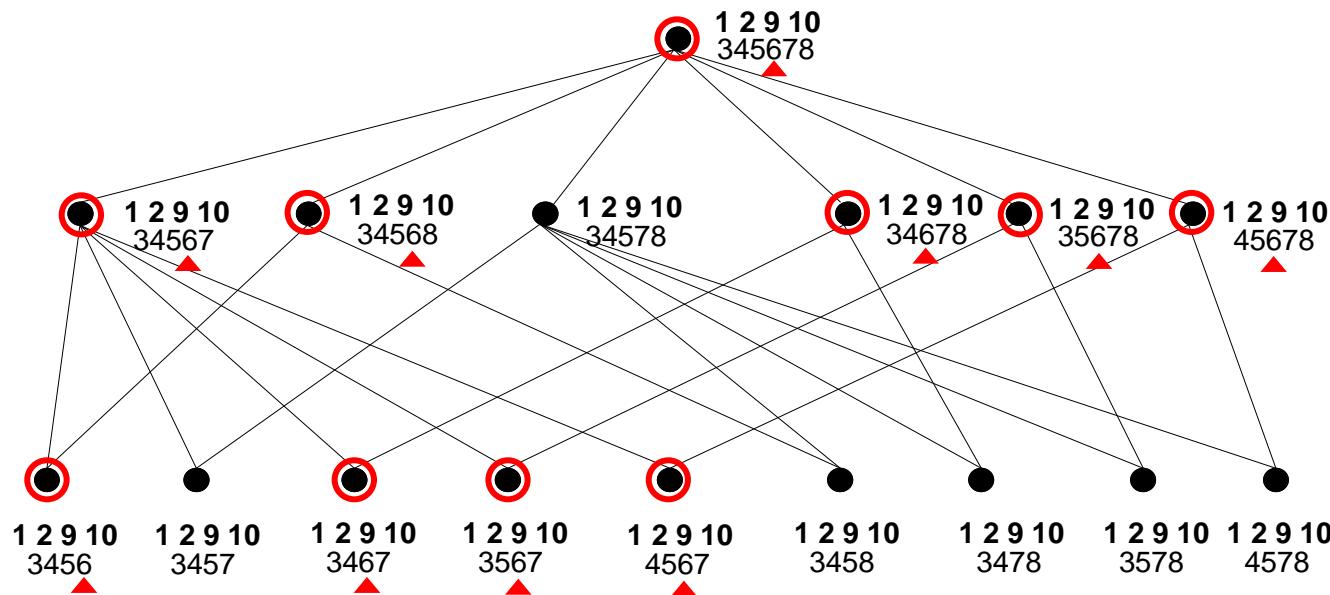
$$W(y_i) = K_i / K$$

$W(y_i) = 1$  for essential sensors

$W(y_i) = 0$  for useless sensors

# **Index of criticity for the sensors**

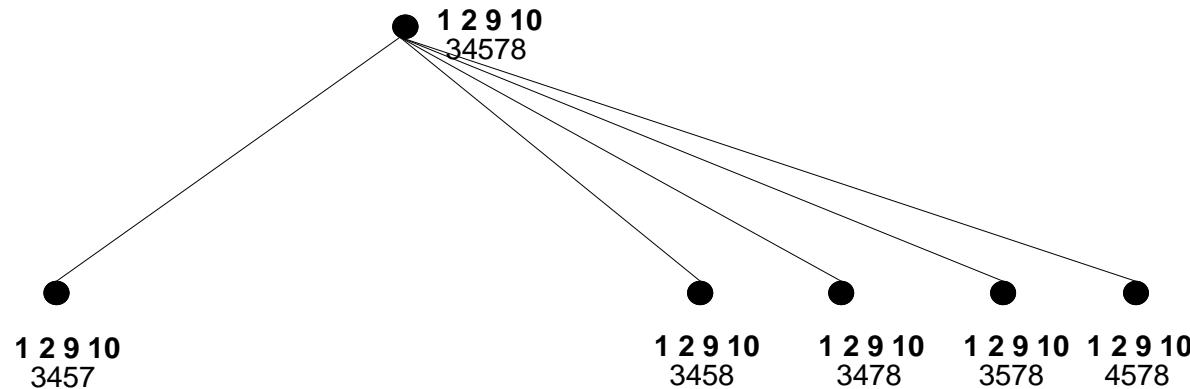
$$D(y_0) = P_0/16$$



# **Index of criticity for the sensors**

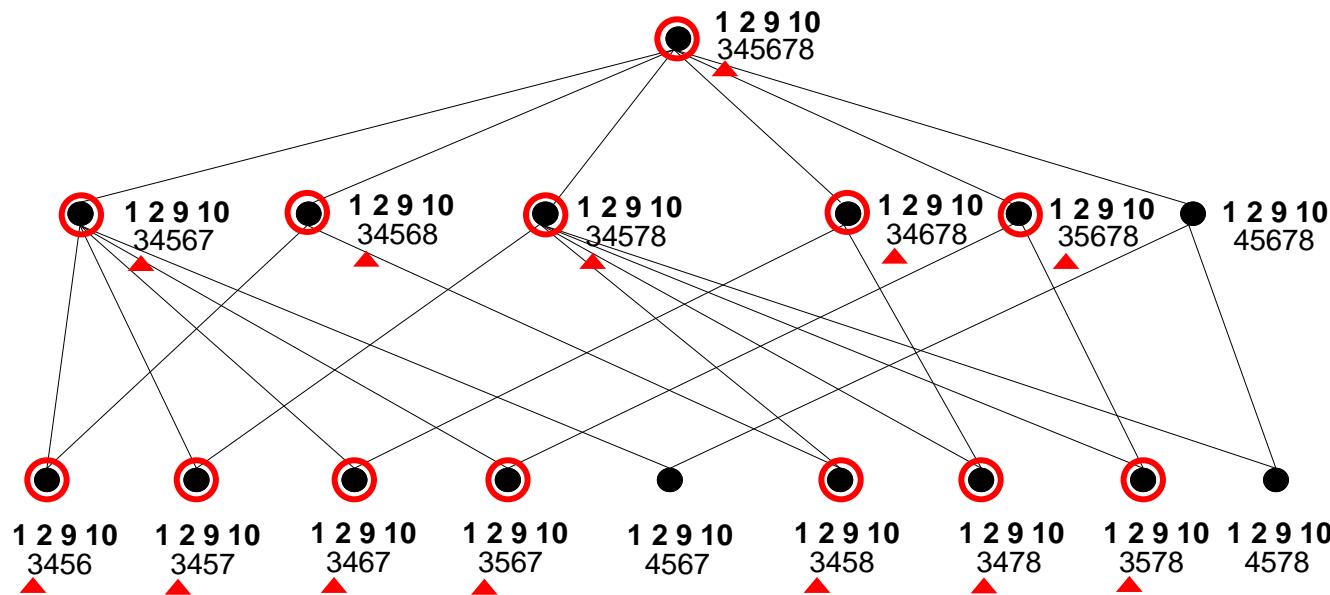
When  $y_6$  is lost:

$$1 - D(y_6) = 6/16$$



# **Index of criticity for the sensors**

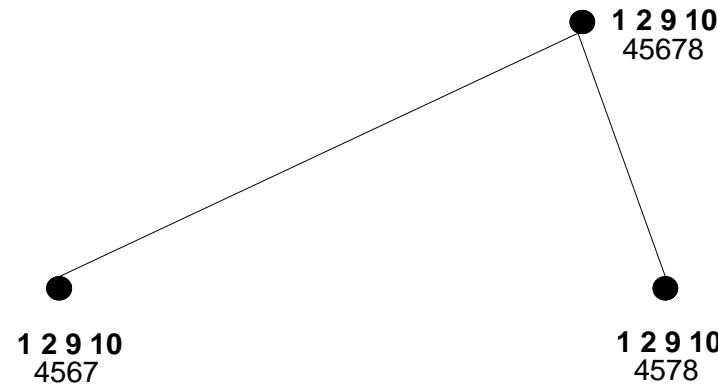
$$D(y_3) = 13/16$$



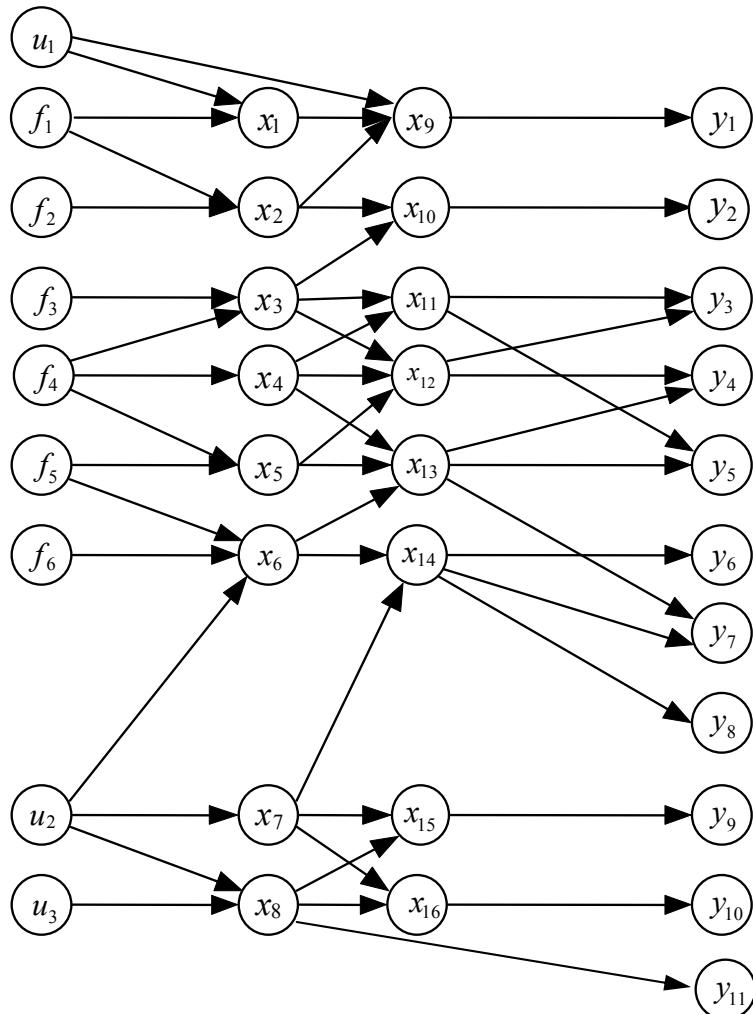
# **Index of criticity for the sensors**

When  $y_3$  is lost:

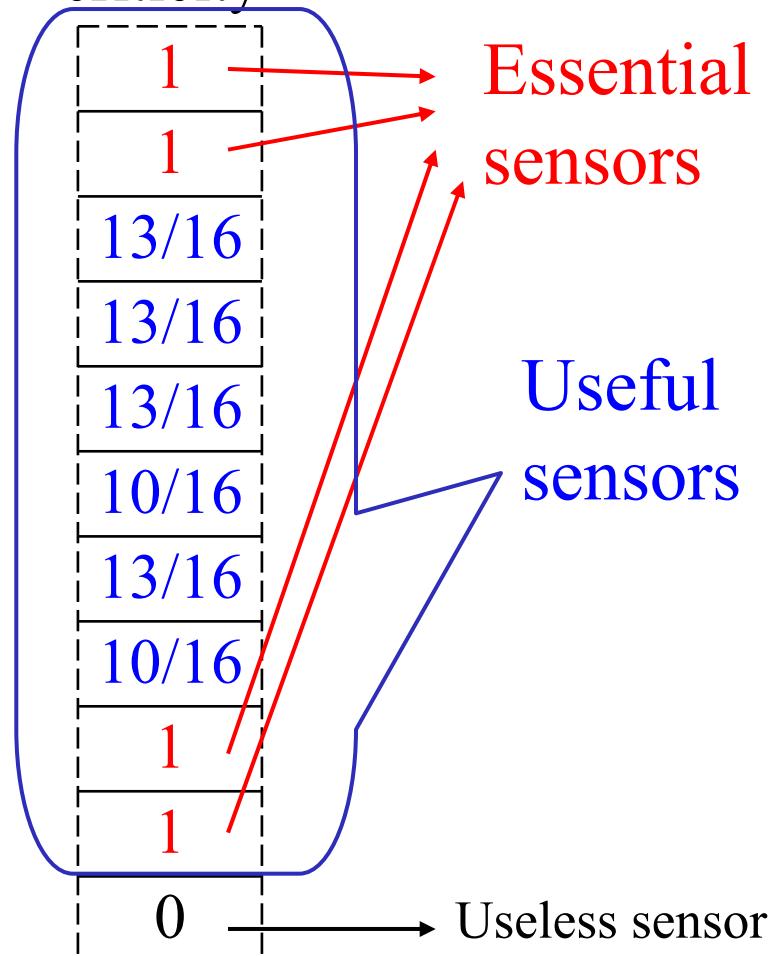
$$1 - D(y_3) = 3/16$$



# Index of criticity for the sensors



Index of  
criticity



# Conclusion

- ❑ Structural modeling of dynamical systems
- ❑ Property preservation under sensor failure for structured systems.
- ❑ Sensor classification with respect to their critical nature concerning FDI.
- ❑ Sensor classification using polynomial time algorithms.
- ❑ Quantitative measure of criticity of sensors
- ❑ This sensor classification can be extended to other problems (disturbance decoupling, ...).

# **Index of criticity for the sensors**

$$W(y_1) = 1$$

$$W(y_2) = 1$$

$$W(y_9) = 1$$

$$W(y_{10}) = 1$$

*Essential*

$$W(y_3) = 13/16$$

$$W(y_4) = 13/16$$

$$W(y_5) = 13/16$$

$$W(y_7) = 13/16$$

$$W(y_6) = 10/16$$

$$W(y_8) = 10/16$$

$$W(y_{11}) = 0$$

*Useless*

# For output connection

$$Y_{cn} = \{y_1, y_2, y_9, y_{10}\}$$

8 minimal sensor sets:

$$\{Y_{cn}, y_3, y_7\}$$

$$\{Y_{cn}, y_3, y_5, y_6\}$$

$$\{Y_{cn}, y_3, y_5, y_8\}$$

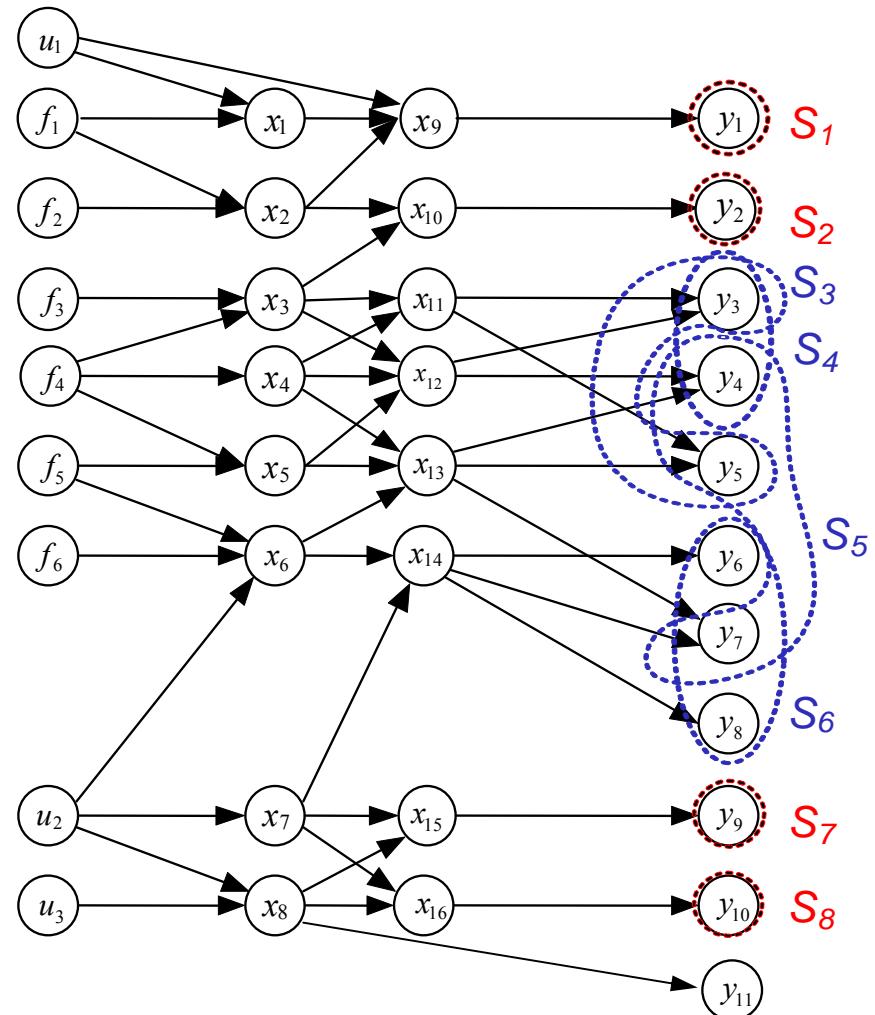
$$\{Y_{cn}, y_4, y_5, y_7\}$$

$$\{Y_{cn}, y_3, y_4, y_6\}$$

$$\{Y_{cn}, y_3, y_4, y_8\}$$

$$\{Y_{cn}, y_4, y_5, y_6\}$$

$$\{Y_{cn}, y_4, y_5, y_8\}$$

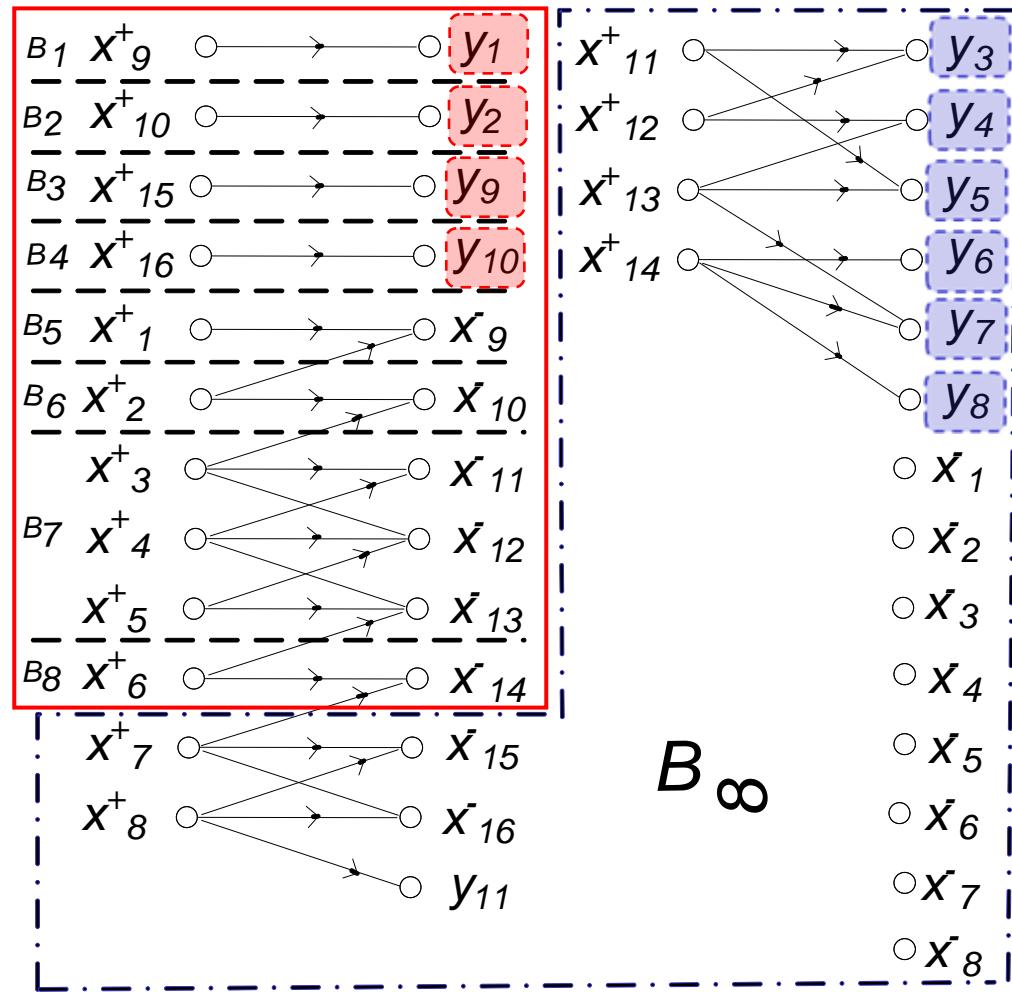


# For contraction avoidance

$$Y_{ct} = \{y_1, y_2, y_9, y_{10}\}$$

9 minimal sensor sets:

- $\{Y_{ct}, y_3, y_4, y_5, y_7\}$   $\{Y_{ct}, y_3, y_4, y_5, y_6\}$
- $\{Y_{ct}, y_3, y_4, y_7, y_6\}$   $\{Y_{ct}, y_3, y_5, y_7, y_6\}$
- $\{Y_{ct}, y_5, y_5, y_7, y_6\}$   $\{Y_{ct}, y_3, y_4, y_5, y_8\}$
- $\{Y_{ct}, y_3, y_4, y_7, y_8\}$   $\{Y_{ct}, y_3, y_5, y_7, y_8\}$
- $\{Y_{ct}, y_4, y_5, y_7, y_8\}$



# FDI rank condition

$$Y_{cf} = \{y_1, y_2\}$$

9 minimal sensor sets:

$$\begin{array}{ll} \{Y_f, y_3, y_4, y_5, y_7\} & \{Y_f, y_3, y_4, y_5, y_6\} \\ \{Y_f, y_3, y_4, y_7, y_6\} & \{Y_f, y_3, y_5, y_7, y_6\} \\ \{Y_f, y_4, y_5, y_7, y_6\} & \{Y_f, y_3, y_4, y_5, y_8\} \\ \{Y_f, y_3, y_4, y_7, y_8\} & \{Y_f, y_3, y_5, y_7, y_8\} \\ \{Y_f, y_4, y_5, y_7, y_8\} & \end{array}$$

