

# Decentralised diagnosis of discrete-event systems: application to telecommunication network

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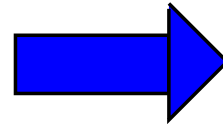


# Telecommunication network

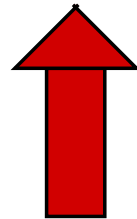
- Set of connected components
  - Distributed system
  - Metropolitan/Wide Area Network
- Purpose: transmission of data between clients (companies)
- Network management: providing a good quality of services
  - “Using all the network resource with a minimal cost”
  - Traffic management, failure management

# Network monitoring

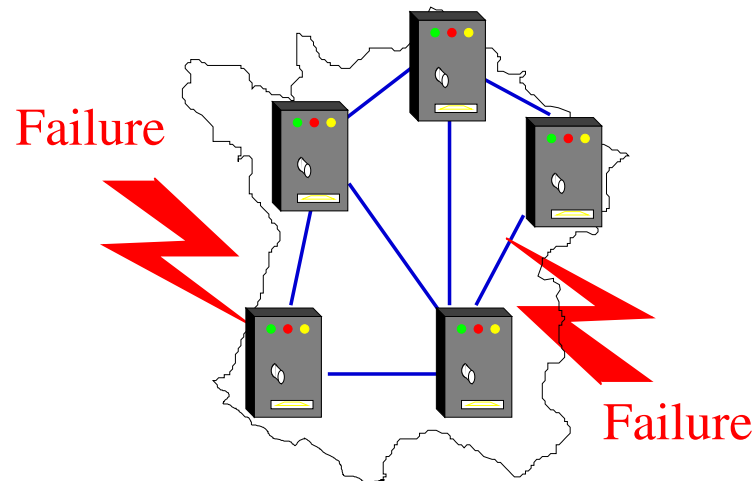
Supervision center



Purpose:  
Identify the problems  
that could have occurred  
which explain the received  
alarms



Alarms



Detection,  
Localisation,  
Identification,  
Propagation

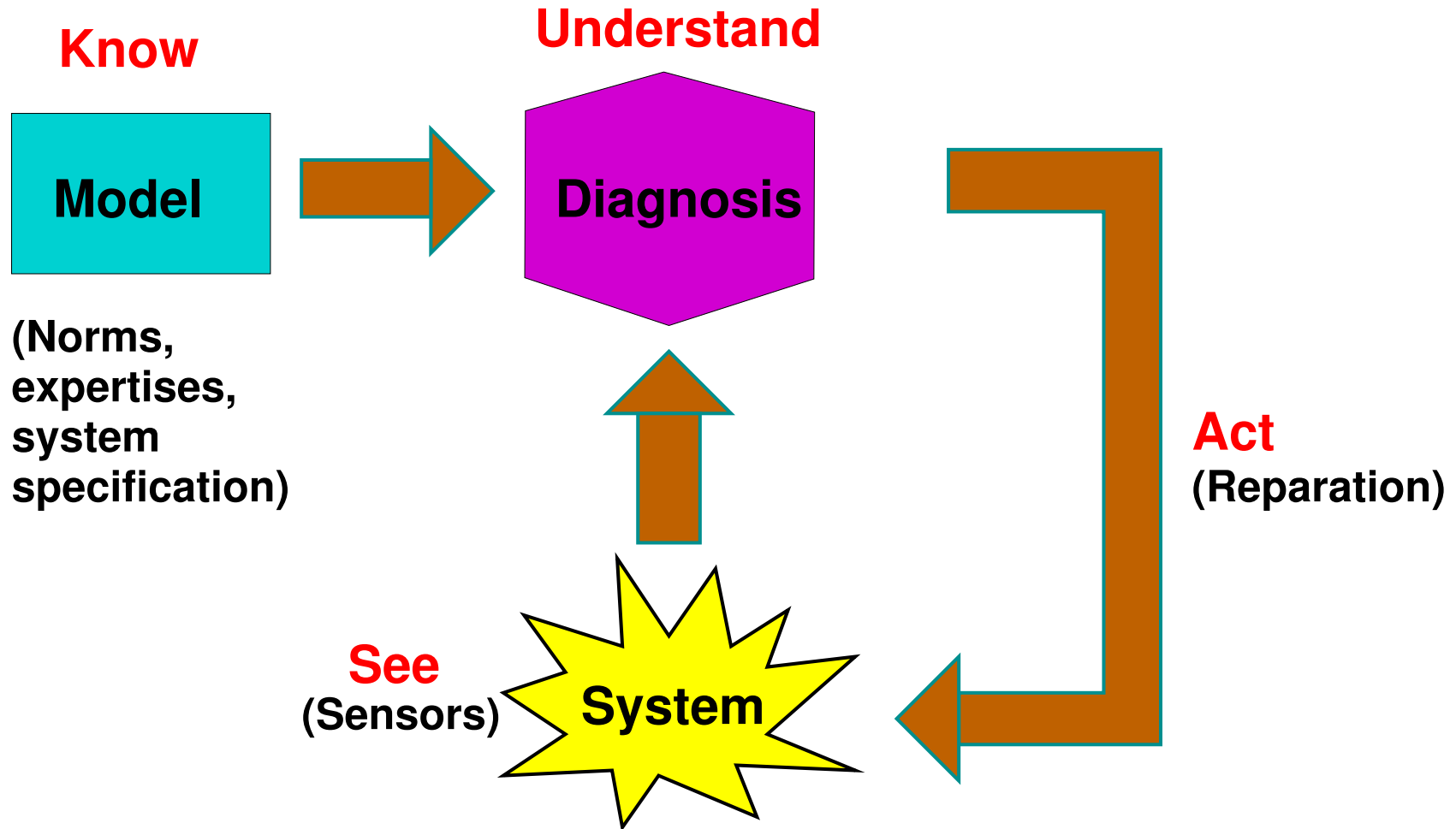
## Needs for monitoring

- Supervised network: large scale system
  - Very important number of received alarms per day (several thousands)
  - Supervisor: human agents
    - \* Analysis of the received alarms: complex problem, in particular if we need to determine the problems quickly
- An automatic system to help in the alarm interpretation is necessary
- Existing systems [Sloman86][Jakobson93]
  - Expert systems, correlation alarm systems
    - \* Problem: evolutive system

## Context and purposes

- To propose a failure diagnosis system
  - Taking into account the evolutivity of the supervised system
  - Producing complete and concise diagnoses
  - Online diagnosis approach: need of efficiency
- Our proposal: decentralised approach
- Context: MAGDA project
  - Academic partners: IRISA, LIPN
  - Industrial partners: Alcatel, France Telecom, Ilog

# Diagnosis: principles



## KNOW: Model

- Model of discrete-event systems
  - *Failure*: occurrence of an event which can change the state of a component
  - *Interaction* between components: message exchanges (emission/reception)
  - *Alarm*: emission of an observable event by a component
- Behaviour of the system
  - Nominal behaviour, Faulty behaviour
- Used formalism:
  - Set of communicating automata

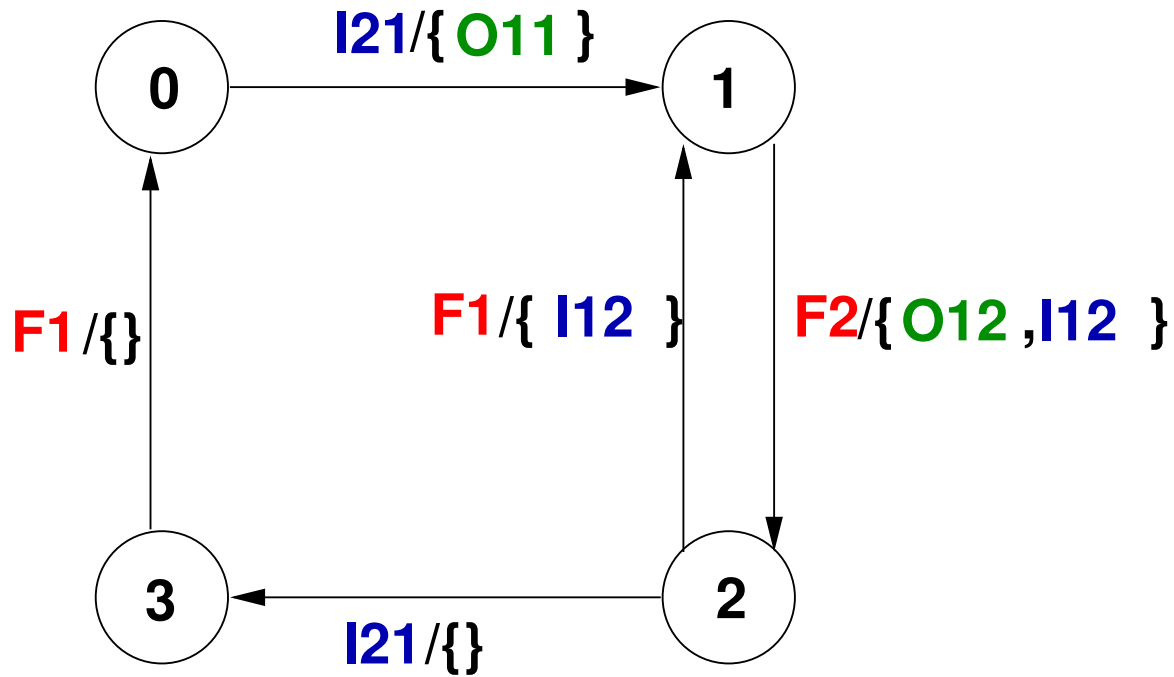
## Model of component

$$\Gamma_i = (Q_i, Exo_i \oplus Rcv_i, Obs_i \oplus Emit_i, T_i)$$

- $Q_i$  finite set of states, modes of behaviour (faulty or not)
- Reception events
  - $Exo_i$  exogenous events: failures, actions from the environment
  - $Rcv_i$  internal events: reception of messages from other components (event propagation)
- Emission events
  - $Emit_i$  internal events: emission of messages to other components (event propagation)
  - $Obs_i$  observable events: emission of alarms to the supervisor
- $T_i \subseteq Q_i \times Exo_i \oplus Rcv_i \times 2^{Obs_i \oplus Emit_i} \times Q_i$  set of transitions



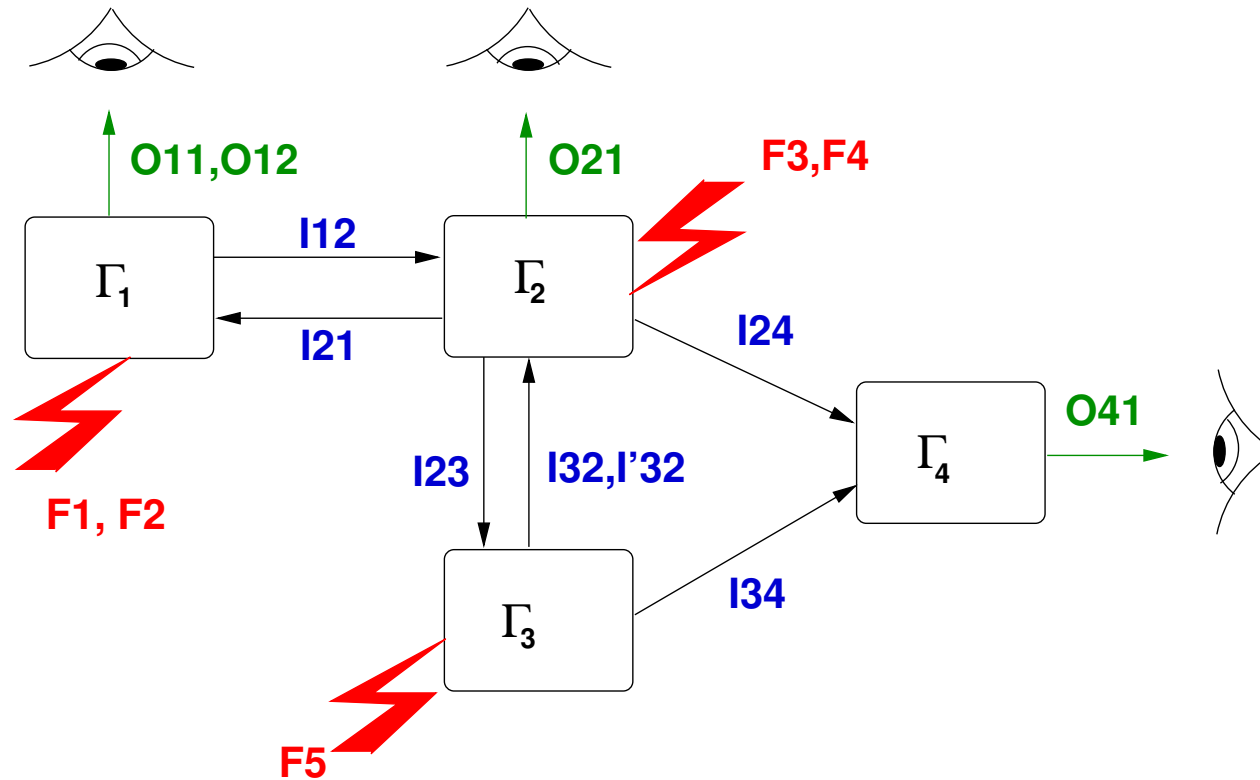
## Model of component: example



- $Exo_i$ : **F1 F2**,  $Rcv_i$ : **I21**
- $Emit_i$ : **I12**,  $Obs_i$ : **O11 O12**

## Model of the system

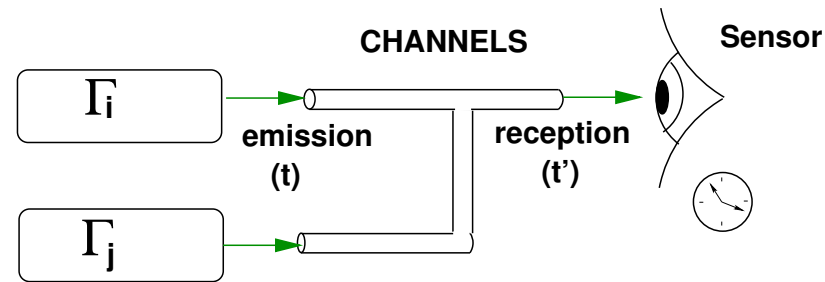
Set of components:  $\Gamma \triangleq \{\Gamma_1, \dots, \Gamma_n\}$



Global behaviour obtained by synchronised product (synchronisation on internal events):  $\|\Gamma\| = \prod_{i \in 1}^n \Gamma_i$

## SEE: Observations

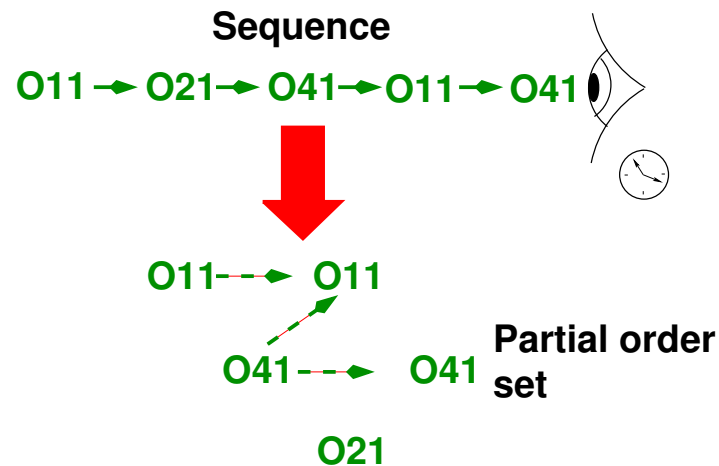
- Set of sensors
  - Observation channels



- Different propagation delays
  - \* Instantaneous
  - \* With a known maximum delay  $D$
- Observation: reception of a message from a component by a sensor
- On a sensor: order of reception  $\neq$  order of emission!
- 2 sensors may not have synchronised clocks!

## Observations: partial order

- $\mathcal{O}$ : set of observations (message with a date of reception)
- $\preceq$ : partial order relation on the observations
  - based on the *observability of the system*
    - \* Number of sensors (synchronised clocks?)
    - \* Characteristics of the channels
      - Instantaneous ? FIFO ? propagation delay ?



## UNDERSTAND: diagnosis

- Purpose: to explain the observations by the occurrence of failures (permanent, intermittent)
  - Given the model  $\Gamma$ , given the observations  $\mathcal{O}$ , how to find the behaviours modelled in  $\Gamma$  that are *compatible* with  $\mathcal{O}$ .
- Diagnosis
  - Set of behaviours
  - Sequences of events that could have occurred on the supervised system
- Diagnosis = Transition system

## Centralised approaches

- *Centralised approaches*  $\Rightarrow$  Need of the *Global Model*

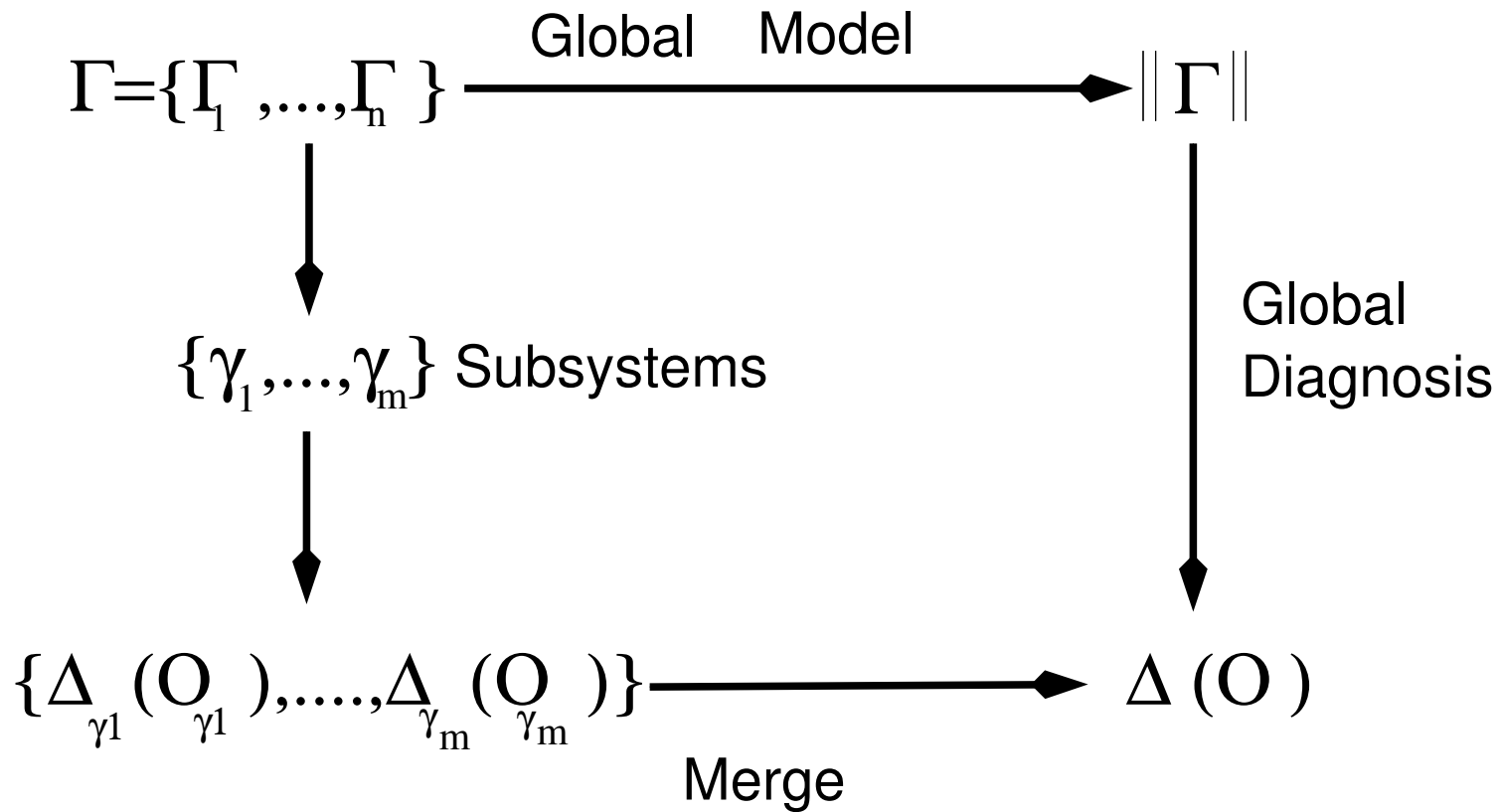
$$\|\Gamma\| = \prod_{i \in 1}^n \Gamma_i$$

- *Diagnoser approach* [Sampath *et al.*][Rozé *et al.*]
  - Based on an *observer*: a finite-state machine which represents the set of *observable behaviours* from the supervised system
  - Diagnosis information: contained in the states of the observer
  - Advantage: the computation of the diagnosis is efficient (parsing of the observer).
  - Drawback: computation of the diagnoser, good luck!
    - \* Worst case size of  $\|\Gamma\|$ :  $\geq 2^n$  (MAGDA project = small network = ( $n = 57$ ))
    - \* Worst case size of *Diagnoser*( $\|\Gamma\|$ ):  $\geq 2^{2^n}$

## Decentralised approaches

- Principle: *Divide and conquer*
  - *Divide*:
    - Computation of a set of *subsystem diagnoses*  $\Delta_{\gamma_1}(\mathcal{O}_{\gamma_1}), \dots, \Delta_{\gamma_m}(\mathcal{O}_{\gamma_m})$ 
      - \* Diagnosis which explains *observations*  $\mathcal{O}_{\gamma_i}$  from a *subsystem*  $\gamma_i = \{\Gamma_{i_1}, \dots, \Gamma_{i_k}\}$  by a set of *subsystem behaviours*
      - \* Explanation based on the hypothesis the subsystem  $\gamma_i$  is independent from the others
  - *Conquer*:
    - Merge of the *subsystem diagnoses* to get the *global diagnosis*
- $$\Delta(\mathcal{O}) = \text{Merge}(\Delta_{\gamma_1}(\mathcal{O}_{\gamma_1}), \dots, \Delta_{\gamma_m}(\mathcal{O}_{\gamma_m}))$$
- Purpose: diagnosed interactions checking

# Centralised/Decentralised





## Advantages of a decentralised approach

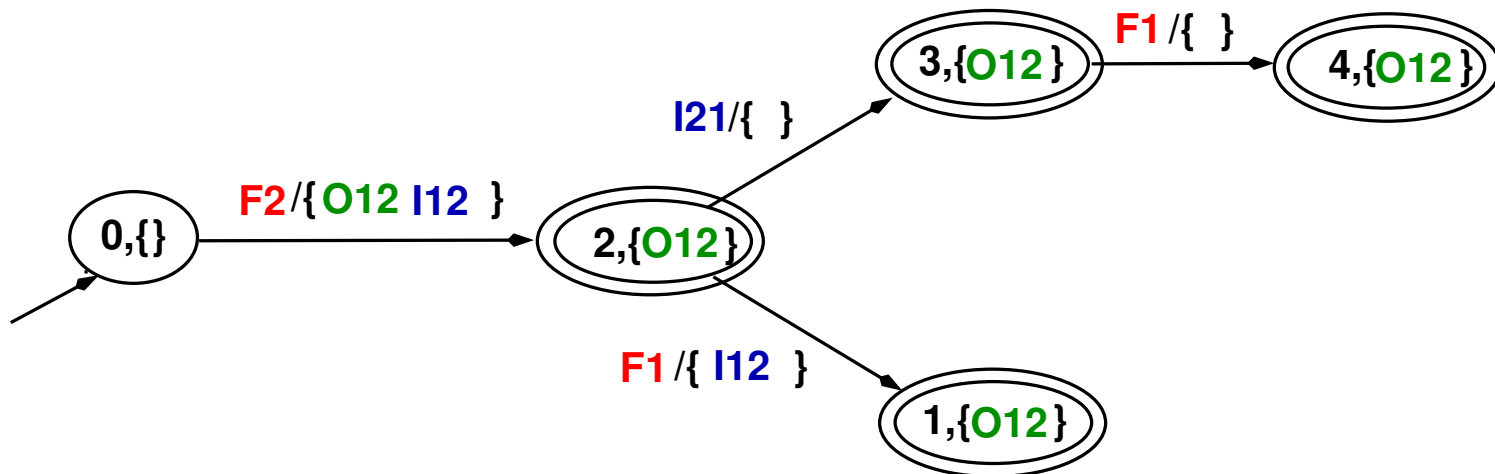
- The *global model* is not necessary
  - Use of *tractable models*
- Supervised systems: distributed systems, well-suited approach
  - More adapted to the evolution, the reconfiguration of a component of the system

## Decentralised approaches: previous works

- Decentralised and coordinated diagnosers [Debouk *et al.*]
  - One diagnoser by sensor: computes a *local diagnosis*
  - Merge: communication protocols between diagnosers to compute the *global diagnosis*
  - Problem: the local diagnosers still need the computation of  $\|\Gamma\|$
- Diagnosis of active systems [Baroni *et al.*]
  - *Simulation* of the decentralised model  $\Gamma$  constrained by the received observations
  - *Simulation* by subsystems (*subsystem diagnosis*), and generalisation of the simulation (*Merge*)
  - Disadvantage: *offline method*, can't be used as a monitoring system (offline diagnosis approach)

## Diagnosis representation

- Diagnosis (subsystem and global)
  - Set of behaviours
    - \* Occurrence of failures and their propagations
- Can be represented by a communicating automaton
- Example: diagnosis of the subsystem  $\gamma_1 = \{\Gamma_1\}$  (observation **O12**)

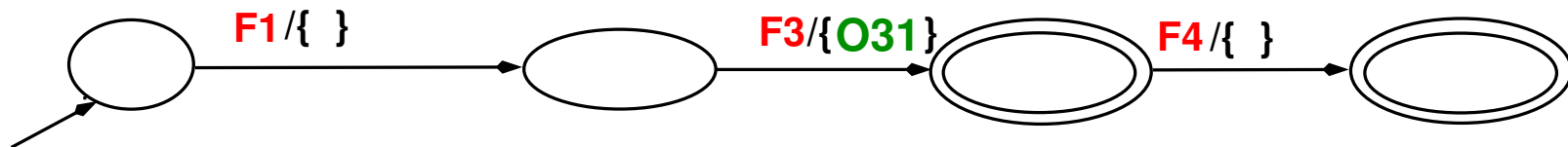


## Reduced representation

- Diagnosis: set of transition paths
  - the number of paths can be important
    - \* essentially due to the concurrency of the system
  - need of a reduced representation
- In the reduced representation,
  - Paths = Event *traces* [Mazurkiewicz 86]
  - equivalent class of *event sequences*
    - \* equivalence based on *event independency* (concurrency)
- Partial order reduction technique

## Reduced representation: example

- If the diagnosis consists of the following sequences
  1.  $F1/\{\}$   $F3/\{O31\}$   $F4/\{\}$
  2.  $F1/\{\}$   $F4/\{\}$   $F3/\{O31\}$
  3.  $F3/\{O31\}$   $F1/\{\}$   $F4/\{\}$
- If we know that  $F1/\{\}$  and  $F3/\{O31\}$  independent,  $F3/\{O31\}$  and  $F4/\{\}$  independent
- The following path is sufficient to represent the diagnosis
  - by successive permutations of consecutive independent events

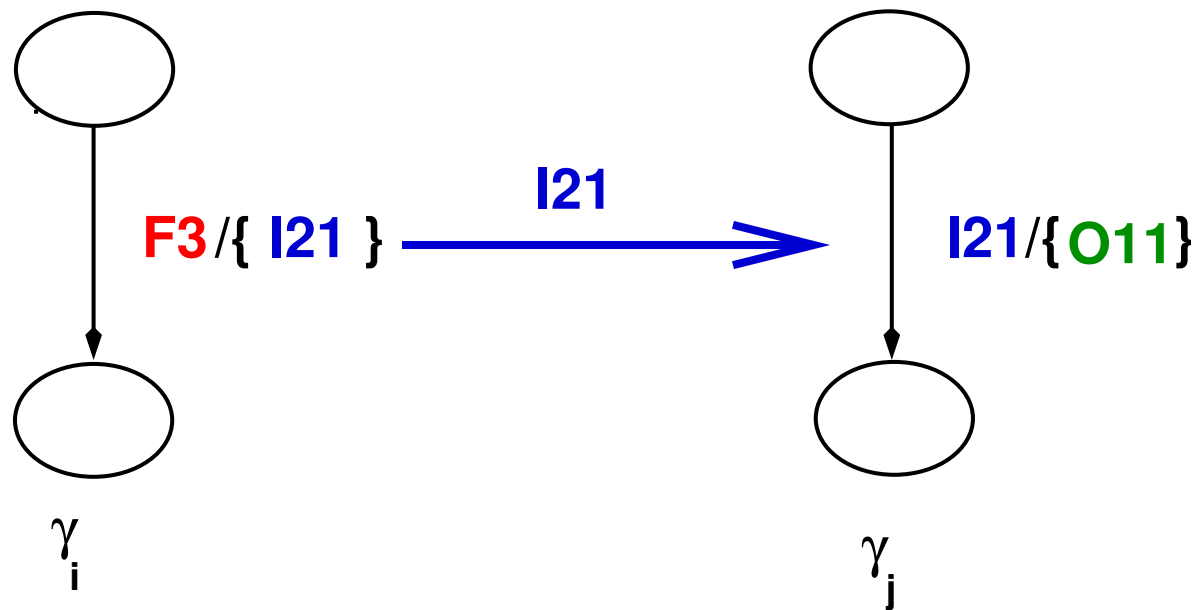


## Local diagnosis computation

- Given a subsystem  $\gamma_i \triangleq \{\Gamma_{i_1}, \dots, \Gamma_{i_k}\}$
- Given  $\mathcal{O}_{\gamma_i}$  the set of observations from  $\gamma_i$
- Purpose: find the set of paths from  $\|\gamma_i\| \triangleq \prod_{j \in \{i_1, \dots, i_k\}} \Gamma_j$  explaining  $\mathcal{O}_{\gamma_i}$
- $\gamma_i$  has been chosen to be tractable
  - A centralised approach can be applied on  $\gamma_i$
  - Use of an adaptation of the diagnoser approach [Sampath *et al.*] in order to have an efficient computation
    - \* Computes paths representing *traces*
    - \* Noted  $\Delta_{\gamma_i}^{red}(\mathcal{O}_{\gamma_i})$ .

## Local diagnoses and interactions

- Local diagnosis of  $\gamma_i$ :
  - Inform about the possible interactions with the *neighbours* of  $\gamma_i$
- Interaction: exchange of events, synchronisation



## Merge operation: characteristics

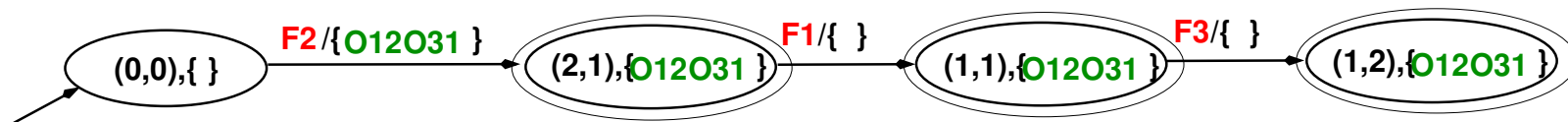
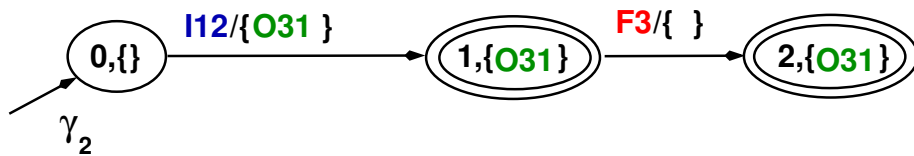
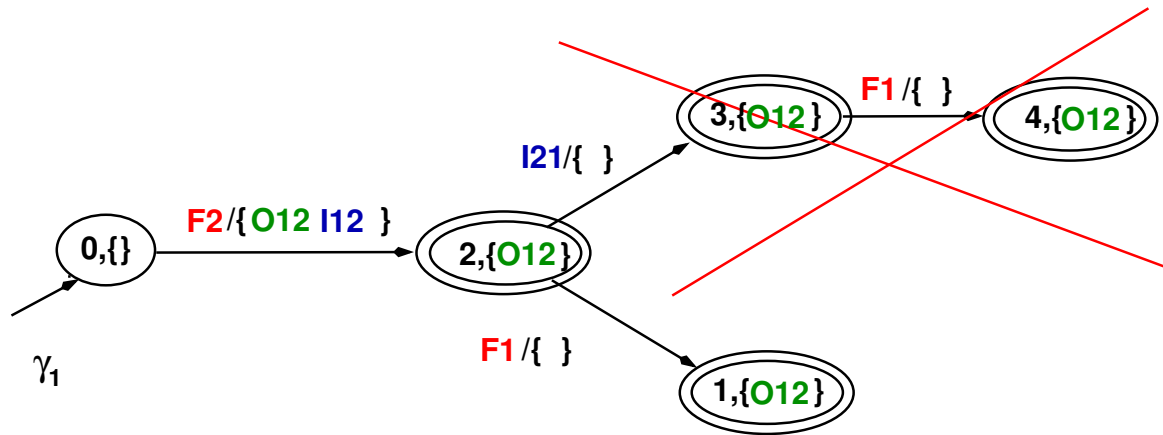
- Purpose: check the interactions between the local diagnoses.
- Merge operation ( $\odot$ ): *synchronised and reduced* product on automata
  - Merge operation recipe:
    1. define a *dependence relation* on the events of the system
    2. mix the results from [Arnold92] (synchronised product of transition system) with the *sleep set algorithm* from [Peled93] (partial order exploration based on the dependence relation) and you have:

$$\Delta_{\gamma_i \cup \gamma_j}^{red}(P_{\gamma_i \cup \gamma_j}(\mathcal{O})) = \Delta_{\gamma_i}^{red}(P_{\gamma_i}(\mathcal{O})) \odot \Delta_{\gamma_j}^{red}(P_{\gamma_j}(\mathcal{O}))$$

where  $P_{\gamma_i}(\mathcal{O})$  is the partial order set of observations extracted from  $\mathcal{O}$  (projection) which have been emitted by  $\gamma_i$



# Merge operation: example



Merge



## Merge strategy

- $\odot$  is based on a product operation, it can be not efficient!
  - we have to use it meanly, when necessary.
- We need a plan for the application of the merge
  - Strategy based on the information contained in the local diagnoses
    - \* What are the diagnoses to merge?
    - \* *The less I merge, the more efficient I am!*
- The strategy is defined with 2 rules
  - Incompatible path detection
  - Selection of dependent diagnoses

## Rule 1: Incompatible trajectory detection

- Let  $E_i$  be the set of exchanged events (interactions) of the subsystem  $\gamma_i$  according to its diagnosis
- Every event  $e$  exchanged between  $\gamma_i$  and  $\gamma_j$  is necessary such that:

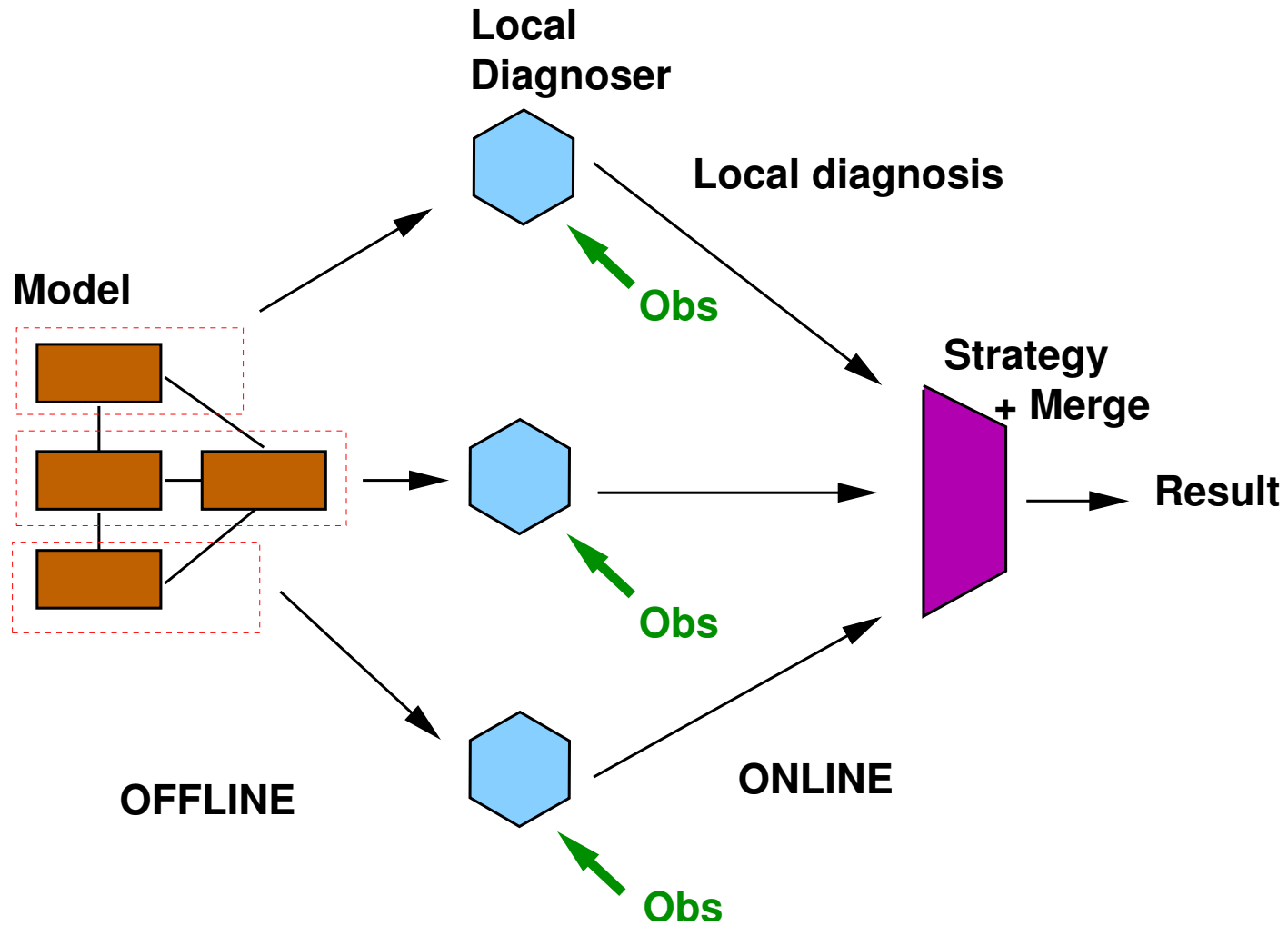
$$e \in E_i \cap E_j$$

- If not, every path containing  $e$  in the diagnosis  $\Delta_{\gamma_i}$  is *incompatible*
- **Rule 1:** elimination of incompatible paths before applying the  $\odot$  operation on the local diagnoses.

## Rule 2: Selection of diagnoses

- Basic idea: merging two diagnoses that do not interact each other
  1. is roughly equivalent to make the Cartesian product
  2. is useless, their interactions are not checked
- **Rule 2:** Only merge diagnoses which interact each other
- It is possible to apply the strategy in a parallel way (distributed application)
- The result is a set of *independent diagnoses*.
  1. Each diagnosis gives the explanations of the observations from a part of the system
  2. There is no interaction between the diagnosed parts

# Summary of the approach

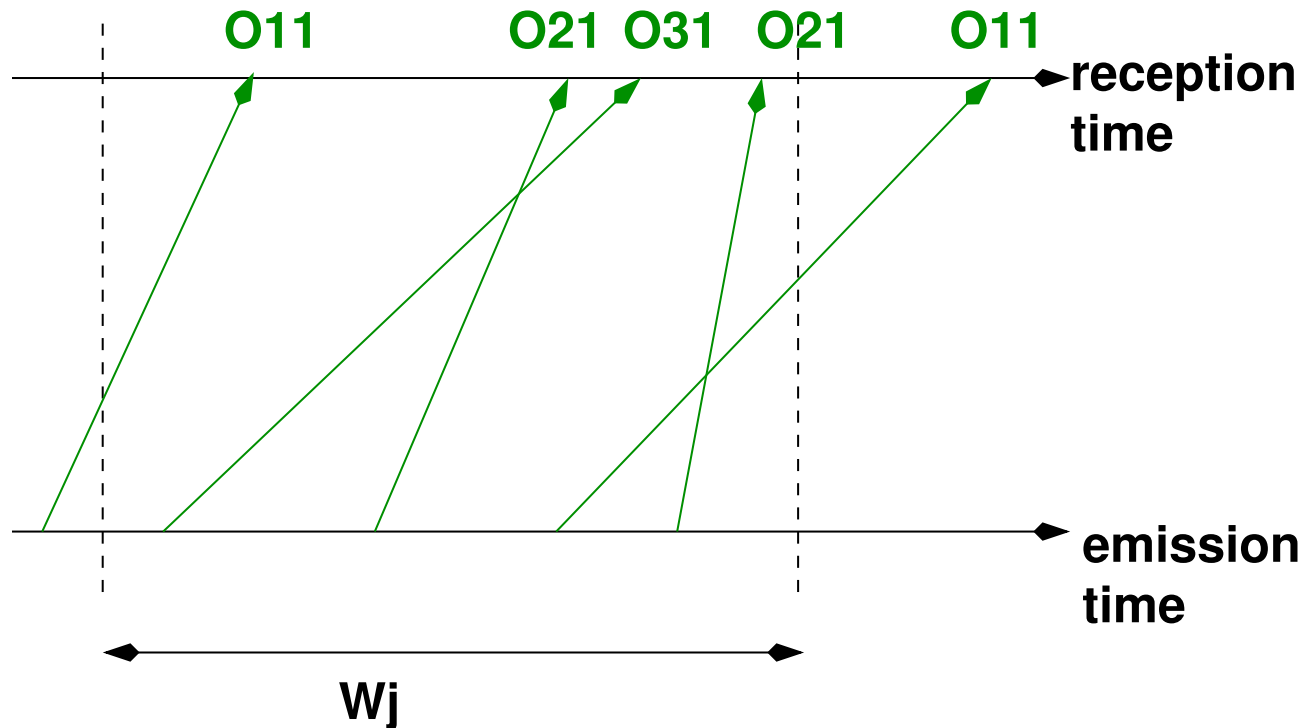


## Incremental diagnosis

- Observations: a continuous flow of alarms
  - Every observation is in a *temporal window*  $W_1, \dots, W_m$
- Given
  - the observations from the window  $W_j$
  - the diagnosis explaining the observations from  $W_1, \dots, W_{j-1}$
- How to efficiently compute the diagnosis explaining the observations from  $W_1, \dots, W_j$ ?
  - Incremental diagnosis computation

## Difficulties

- Generally, at the end of any temporal window, we do not have the guarantee that the received observations can be explained! Some observations might be missing



**O11 is not received during  $W_j$  but can be necessary to make a diagnosis**  
**The sequence O11 O21 O31 O21 may have no explanation**

## First solution: sound temporal windows

- Detection of *sound* temporal windows
  - An observation is emitted and received in the same window
- Can be detected relying on the observation channel properties
- Incremental diagnosis computation:
  1. From the current diagnosis states at the end of  $W_{j-1}$
  2. Computation of the global diagnosis explaining the observations from  $W_j$
  3. Refinement algorithm:  $\Delta_{1,\dots,j} \triangleq \Delta_{1,\dots,j-1} \oplus \Delta_{W_j}$



## General solution

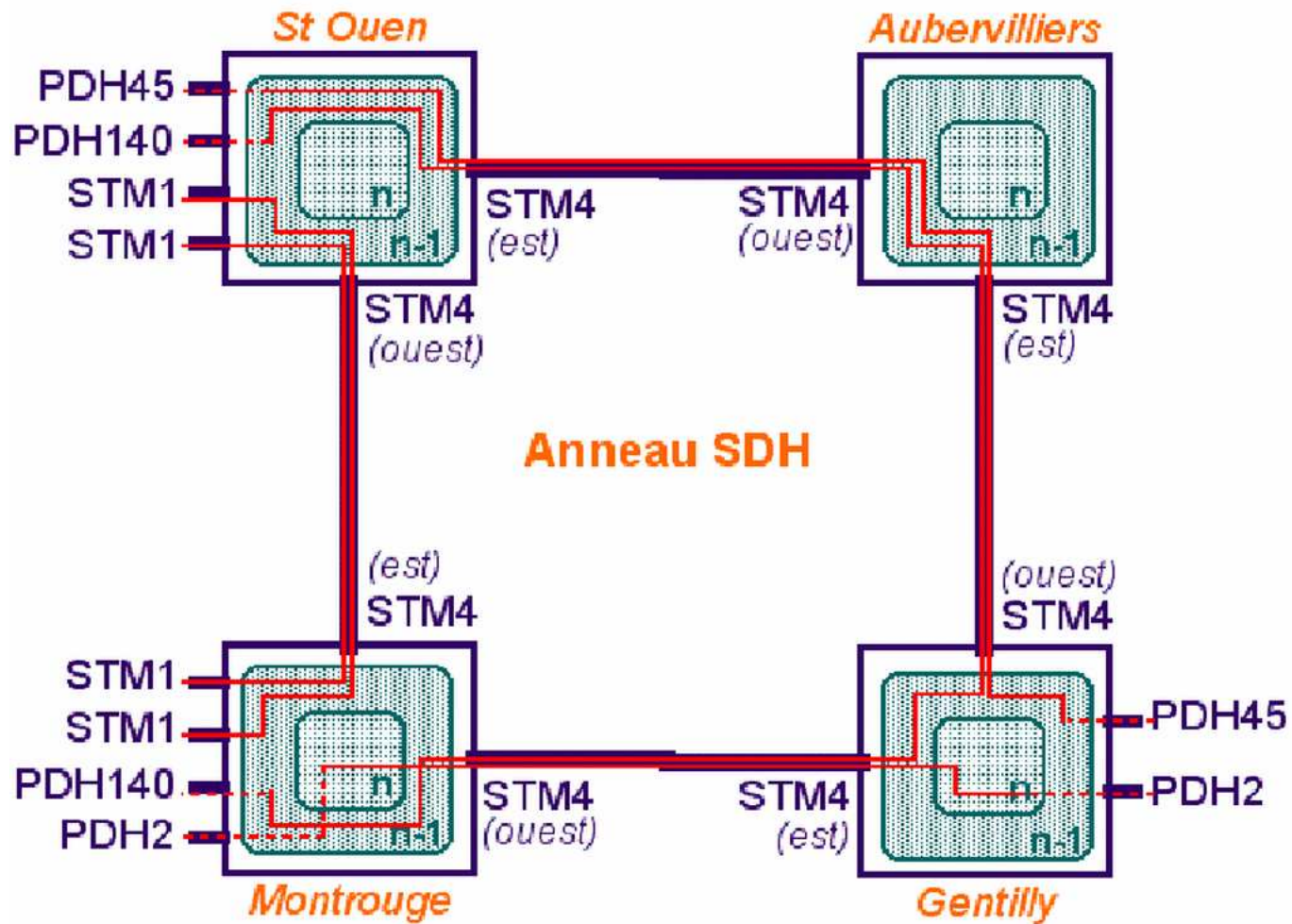
- In the worst case, there is no sound temporal window!
- Need of an *extended diagnosis*  $\Delta_{W_j}^{ext}$ 
  - explains the observations of  $W_j$
  - and some *hypothetical* observations, emitted before the end of  $W_j$  but not received yet
  - has more explanations than the *real* one
- Incremental diagnosis computation: same algorithm as before
- We have the guarantee that if  $W_j$  is sound

$$\Delta_{1\dots j}^{ext} = \Delta_{1\dots j}$$

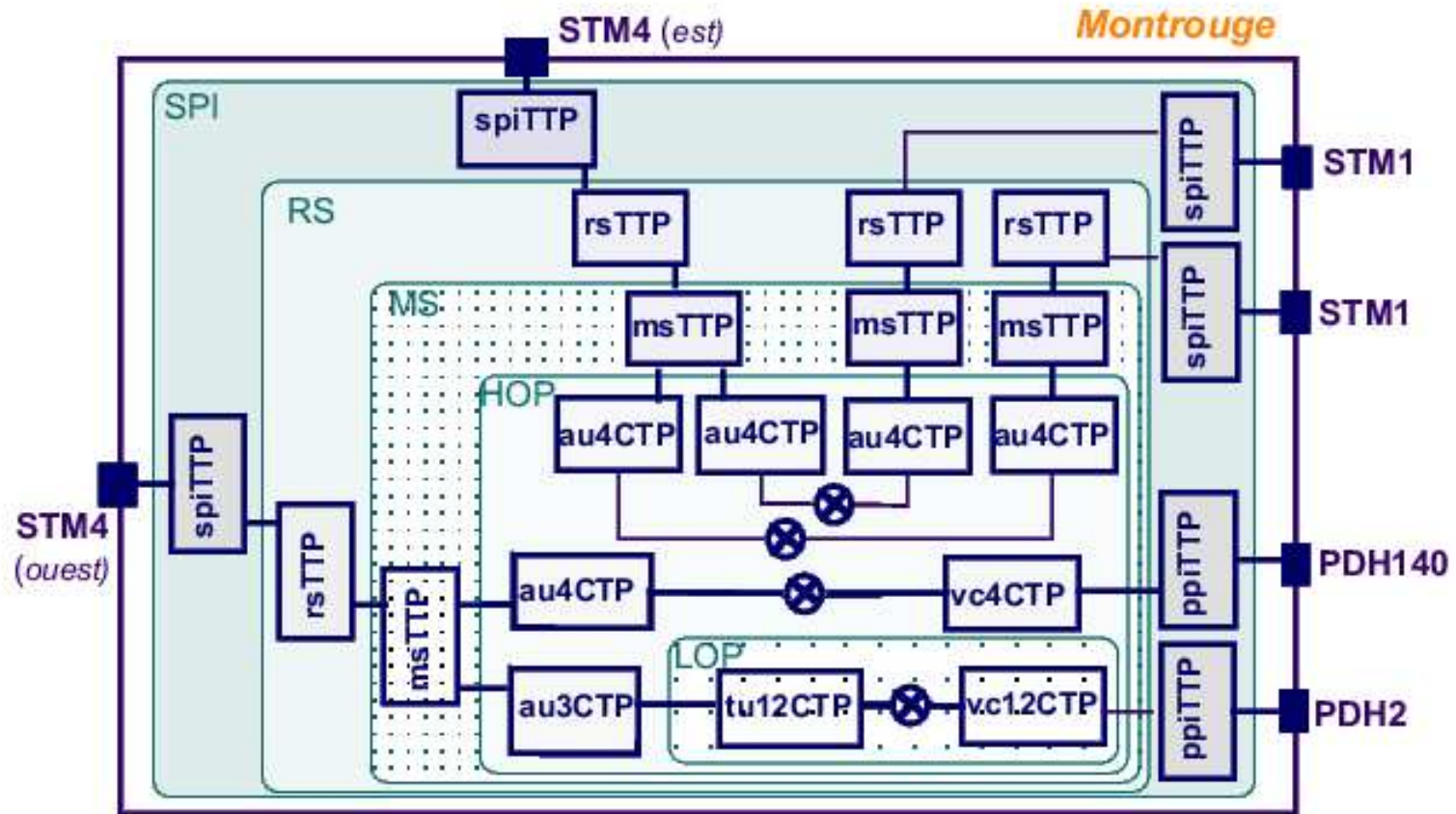
## Transpac network

- French packet switching network
- Experiments done on a sub-part of the network
  - 8 switches, 32 control stations, 2 technical centers
  - Diagnosis difficulty: masking phenomenon
- One studied scenario with 56 alarms
  - Multiple faults diagnosis (masking phenomenon)
  - Result obtained in 8 seconds

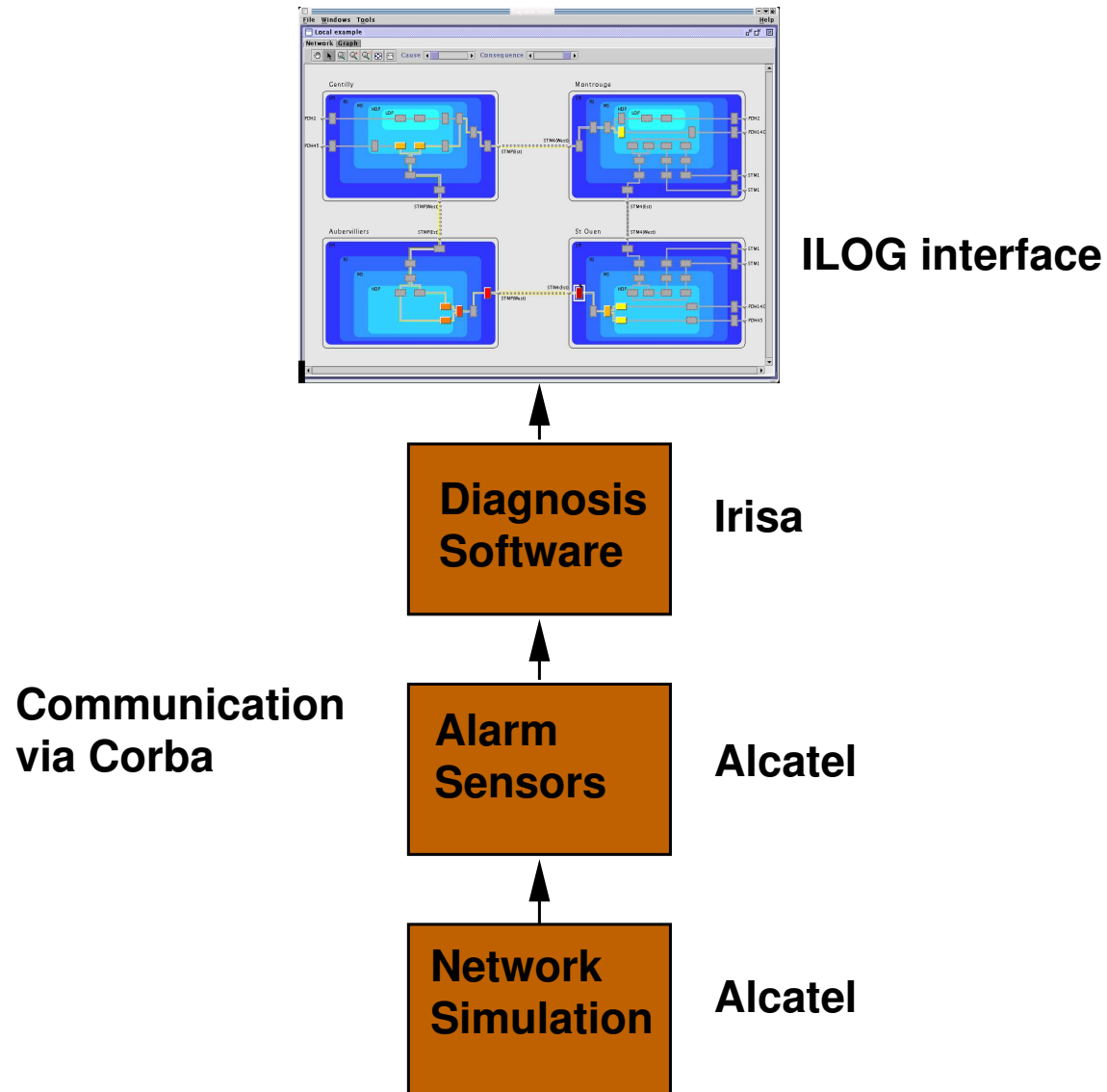
# Magda project: SDH network



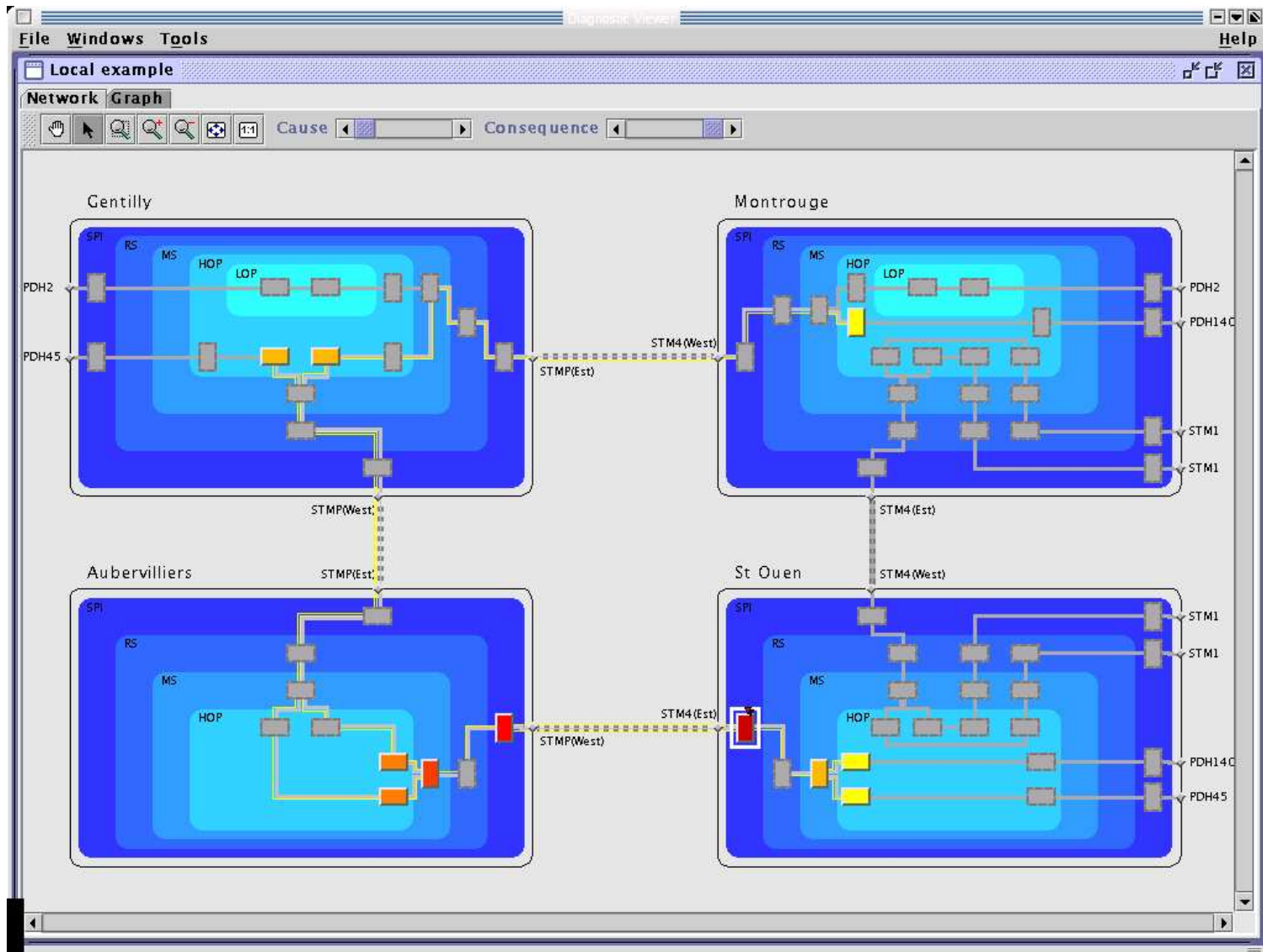
# Magda project: Montrouge ADM



# Magda project: Supervision chain (November 2001)



# Magda project: Interface



## Magda project: studied scenarios

Scenarios	Strategy 1	Strategy 2	Strategy 3	Strategy 4
1	3s 590ms	4s 200ms	16s 540ms	>5mn
2	1s 300ms	1s 300ms	1mn 52s 770ms	>5mn
3	1s 780ms	1s 910ms	>5mn	>5mn
4	1s 600ms	2s 30ms	49s 120ms	>5mn
5	2s 620ms	5s 500ms	5s 430ms	3mn 45s 600ms
6	1s 780ms	2s 320ms	24s 240ms	57s 440ms
7	1s 480ms	1s 700ms	2mn 54s 920ms	>5mn
8	1s 830ms	3s 90ms	3s 30ms	>5mn

- Eight studied scenarios
  - Strategy 1: The previously described strategy
  - Strategy 2: Perturbation of the order of merging
  - Strategy 3: Same as 1 without incompatible path elimination
  - Strategy 4: Same as 2 without incompatible path elimination

## Conclusions

- What a funny challenge it was!!
- Main problem: the use of centralised approaches impossible
  - Large scale DES: problem of spatial complexity
- Framework of a decentralised diagnosis approach
  - “Divide and conquer” principle
    - \* Transfer of a part of spatial complexity to temporal complexity
    - \* In practice, the number of behaviours explaining a set of observations is very small compared to the number of behaviours of the system
  - “Conquer”
    - \* Need of merging strategies,
    - \* Use of diagnosis trace representatives
    - \* Incremental algorithms



## Perspectives

- How to take benefits from the symbolic representation techniques inside this framework? (BDDs)
- How about the diagnosability test of such systems?
- How about using diagnosability for making diagnosis abstractions?
- How to take into account reconfigurations of the system?
- How to mix with planning approaches (repairing plans)?
  - Large scale autonomous systems