Petri Net model-based distributed diagnosis

George Jiroveanu and René Boel EESA-SYSTeMS research Group Ghent University

goal

- On-line Model based fault detection and identification based on observation of occurrence of some events (alarms, messages,...)
- Computationally efficient algorithms for large Discrete Event Systems

modelled as interacting Petri net components

 Using communication between local agents of individual components

goal

Intended applications: Power system backup protection Incident detection in road traffic network

Extensions?

Methodology should allow easy extension to

- time(d) Petri net models
- probabilistic DES models

outline

- Petri net models of interacting components
- Distributed diagnosis with communication between local agents
- Methodology: Backward search to generate minimal explanations
- Distributed diagnosis algorithm
- Conclusions and extensions

Petri net modelling

- Petri net encodes the constraints on the evolution of discrete event systems
- This statement is equivalent to:
 Petri net with given initial state defines
 language = set of allowable traces of events
- "state" encoded by "marking of places"

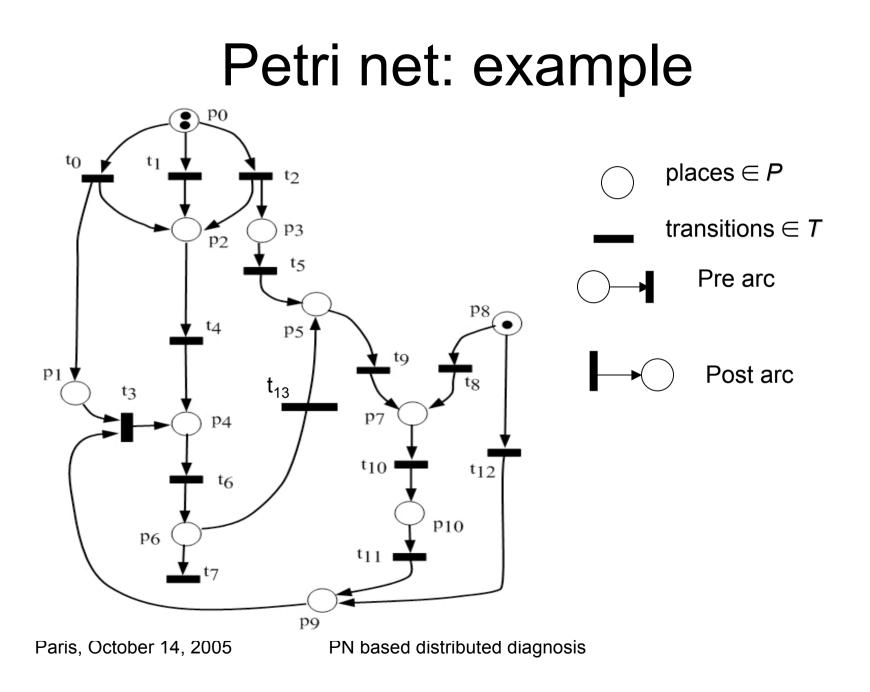
Petri net definition

• structure: a Petri net M = (P, T, Pre, Post)

where *P* = set of places

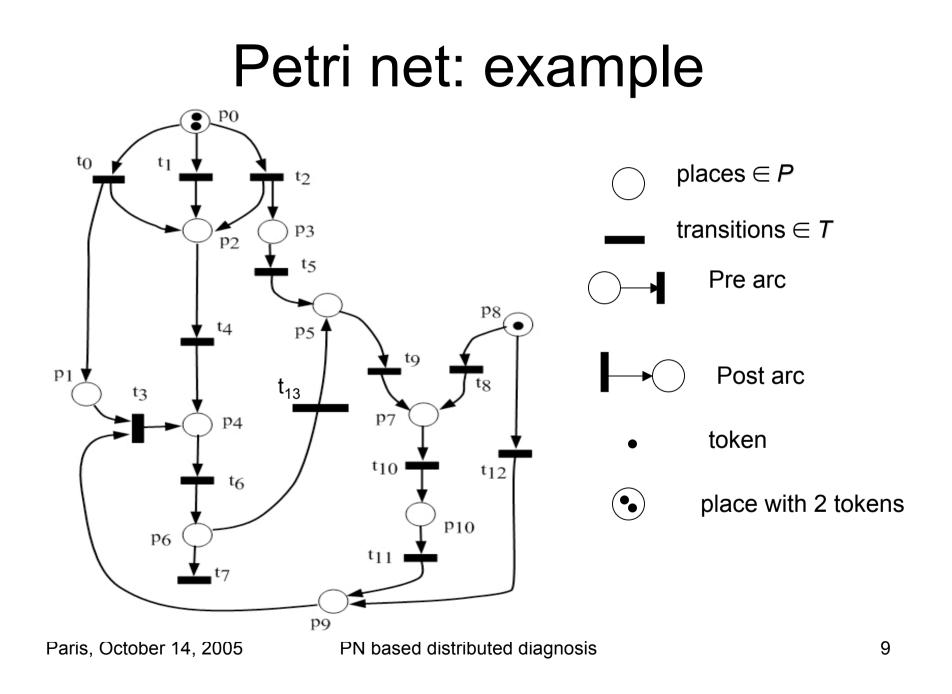
T = set of transitions

- Pre = mapping from P x T into {0, 1}
 - = incidence function defining arcs from place to transition
- Post = mapping from TxP into {0, 1}
 - = incidence function defining arcs from transition to place

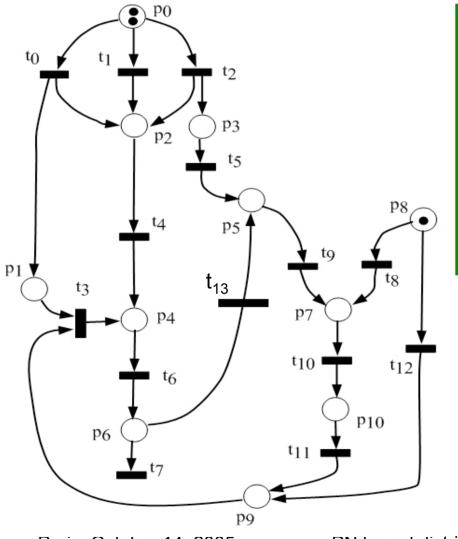


Petri net: marking (state)

 Marking m of Petri net № P → N is integer-valude # P-vector assigning to each place p a (natural) number m(p) of tokens



Petri net: dynamic behaviour

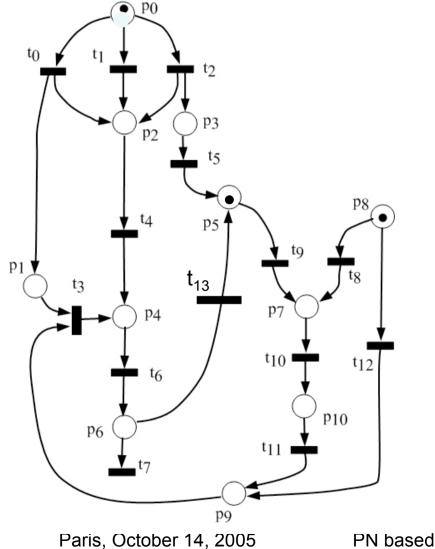


enabling rule: transition t is enabled if ∀p: Pre(p,t) ≤ m(p) if all input places of t contain at least 1 token

in example transitions $t_0, t_1, t_2, t_8, t_{12}$ are enabled

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Petri net: dynamic behaviour



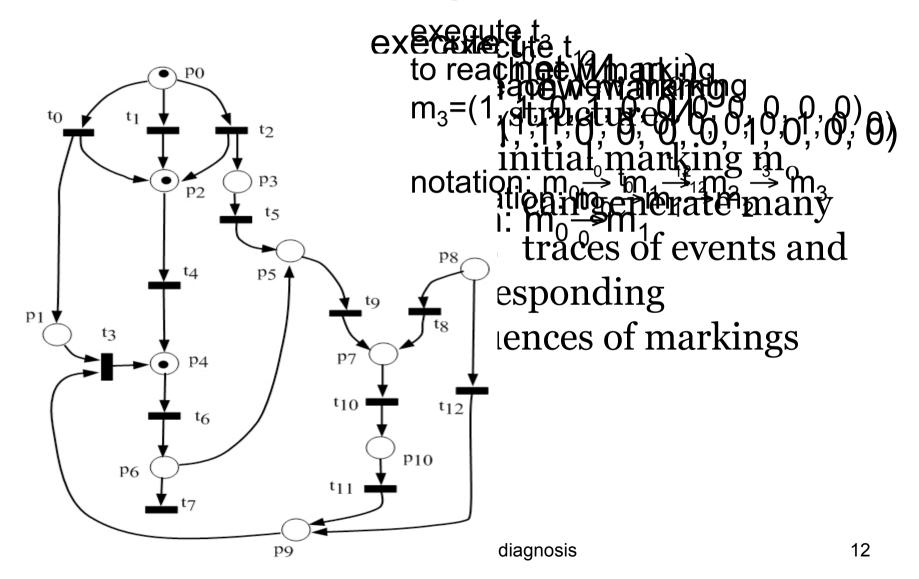
if t is enabled then t can fire

firing rule: when t fires new marking m' = m - Pre(.,t) + Post(t,.) is produced

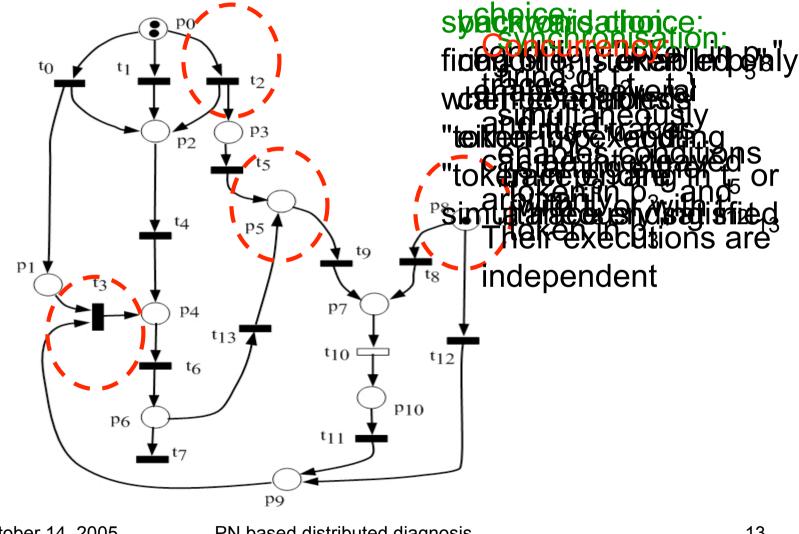
> in example when enabled transition t_2 fires the new marking m' puts 1 token in p_0 , p_3 , p_8

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Petri net: example of evolution



Petri net: modelling power



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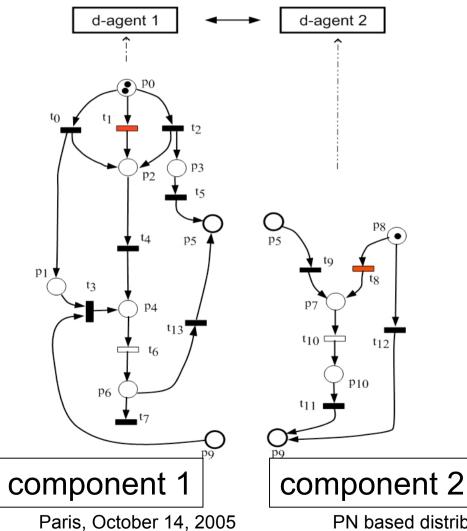
Petri net: allowable traces, languages and reachable sets

- a trace $\tau = m_0 \stackrel{t_1}{\rightarrow} m_1 \stackrel{t_2}{\rightarrow} m_2 \stackrel{t_3}{\rightarrow} m_3 \stackrel{t_4}{\rightarrow} \dots \stackrel{t_{N-1}}{\rightarrow} m_N$ is allowable if the firing condition of the successive events is always satisfied
- The language L_N(m₀) generated by the Petri net N with initial marking m₀ is the set of all allowable traces τ

Petri net: allowable traces, languages and reachable sets

- The language L_N(m₀) generated by the Petri net N with intial marking m₀ is the set of all allowable traces (depending on the context markings are included in language, or only sequence of events is considered)
- The reachable set $\Re_{\mathbb{N}}(m_0)$ is the set of all markings included $\mathscr{L}_{\mathbb{N}}(m_0)$

Petri net: compositional modelling



Large plants

can be represented by several Petri net components, interacting with each other by exchanging tokens via common places

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Petri net: compositional modelling

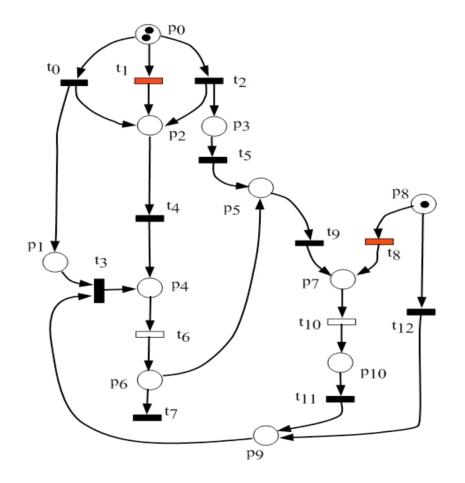
- assumptions
 - -set *T* of transitions partitioned
 - -set *P* of places consists of
 - "local places" in each component i
 - for component i: "input places $P_{IN,i,j}$ that have input transitions (Pre) in component j and output transitions (Post) in component i
 - for component i: "output places $P_{OUT,i,j}$ that have output transitions (Pre) in component j and input transitions (Post) in component i

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Petri net: compositional modelling

- To simplify presentation assume overall Petri net bounded, i.e. all reachable markings have bounded number of tokens in each place
- Difficulty: this assumption depends on the global structure of the Petri ent, cannot be verified locally!

Petri nets: observable events



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Petri net: projection operator

- Observed sequences of events are obtained form an allowable sequence $\boldsymbol{\tau}$
- by the projection operator \square :
 - $-\prod(\varepsilon) = \varepsilon$ where ε is the empty string
 - $-\prod(t) = t$ if $t \in T_o$ where T_o is the set of observable transitions
 - $-\prod(t) = \varepsilon$ if $t \in T_o$ where T_{uo} is the set of unobservable transitions

 $-\prod(s.t) = \prod(s). \prod(t) \text{ for } s \in \mathcal{L}_{\mathbb{N}}(m_0) \text{ and } t \in T$

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Outline

- Petri net model
- model-based diagnosis:
 - problem formulation in centralised case
 - problem formulation in distributed case
 - minimal explanations

Given Petri net \mathbb{M} with known initial marking m₀, after observing a sequence

 $OBS = t_1^{\circ} \cdot t_2^{\circ} \cdot \cdot \cdot t_K^{\circ} \in T_o^*$ of observable events

determine:

the set $\mathcal{E}(OBS)$ of all explanations $\mathcal{E}(OBS) = \{ \text{allowable traces } \tau \in \mathcal{L}_{\mathbb{N}}(m_0) \}$ such that the projection $\prod(\tau) = OBS \}$

- all traces τ in $\mathcal{E}(OBS)$ explain OBS in the sense that τ
 - contains the observed events in the proper ordering,
 - and τ contains no other observable events
- Denote by $\mathcal{M}(OBS)$ the set of all markings reachable in Petri net (\mathcal{M} , m_0) by executing a trace τ in $\mathcal{E}(OBS)$

- Denote by $\mathcal{D}(OBS)$ the set of all faults t_f in any trace τ in $\mathcal{E}(OBS)$
- when $\mathcal{D}(OBS)$ is empty the plant is in the normal state,
- when all traces in $\mathcal{D}(OBS)$ contain at least one fault event t_f then the plant is faulty
- when some traces in $\mathcal{D}(OBS)$, but not all of them, contain a fault event t_f then the plant state is uncertain Paris, October 14, 2005 PN based distributed diagnosis

- assume plant model *I* known
- assume initial marking m₀ known
- remember assumption: set of reachable markings of the Petri net is bounded
- then it is in principle possible to diagnose the plant via forward reachability analysis:

Model based diagnosis via forward reachability analysis

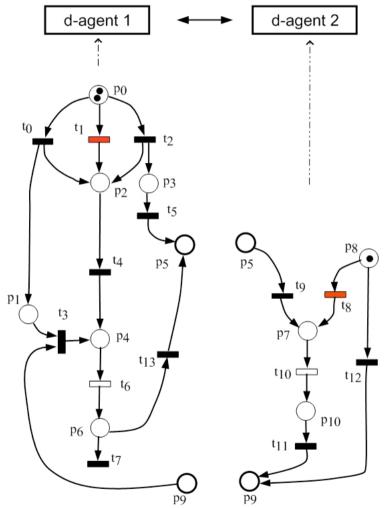
- Calculate (enumerate) all allowable traces $\tau \in \mathcal{L}_{\mathbb{N}}(\mathsf{m}_0)$ and their projection $\prod(\tau)$
- Preserve all allowable traces s.t. $\prod(\tau) = OBS$
- This evaluates the set ε(*OBS*) and allows diagnosis

Sampath, Lafortune et al. (1995) developed an efficient algorithm to implement this approach in an on-line fashion (constructing an observer automaton taking observations as inputs) Paris, October 14, 2005 PN based distributed diagnosis 26

Model based diagnosis via observer automaton

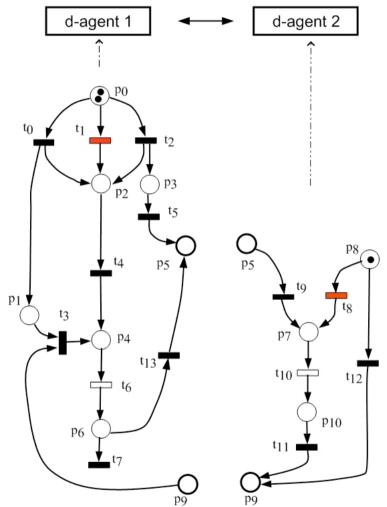
- Recursive (on-line) analysis avoids enumeration of traces that are incompatible with past observation:
- each new observation adds further constraints to set of explanations
- BUT: state space of observer automaton grows in the worst case exponentially in size of Petri net *N*

- For large plant models observer automaton becomes very large
- For large plants it is often difficult to guarantee that all plant updates (modifications) are always known by central agent
- Sensors communication to central node may be unreliable (lossy link or communication delay)

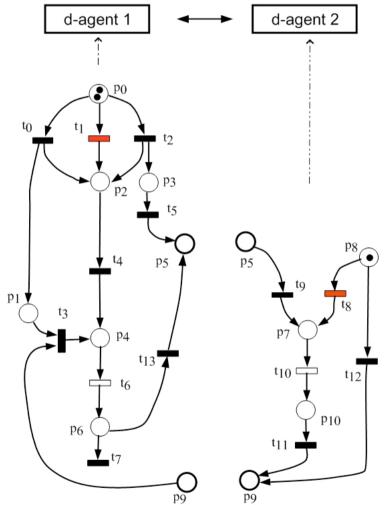


Each Month Plant has needed ting local period with the second sec knowing local model only knowing what places it has in common with which neighbouring component receiving local sensor output exchanging message with neighbouring d-agents

- Forward reachability method not applicable
 - agents do not know the complete initial condition of the local component since they do not know when tokens can arrive from neighbouring agent
 - apart from computational problem of exponential growth of observer automaton



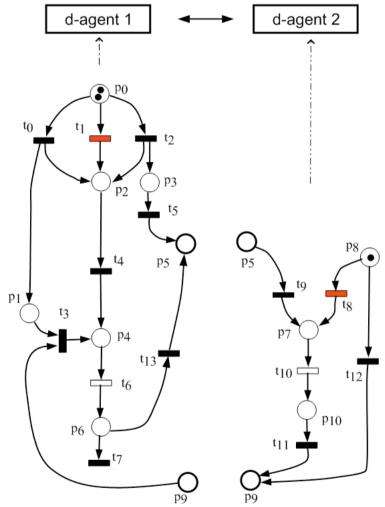
Each local d-agent •knows local model and local initial condition •knows places in *P*_{*IN,i,j*} where it can unpredictably receive tokens from neighbouring component j



Each local d-agent receives local observations (projection OBS_i of OBS onto locally observable tranistions in $T_i \cap T_o$)

Local agent tries to calculate set of local explanations (projections of allowable traces onto local transitions) for OBS_i

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Each local d-agent i CANNOT enumerate all local explanations, without a lot of unnecessary calculations since it would have to assume that each input place in P_{INII} contains a number of tokens specified by the a priori upper bound (which requires global calculation anyway)

- Most previous research on FDI (Benveniste, Fabre, et al.; Genc and Lafortune; Lamperti et al.; Su and Wonham;...)
- resolves this problem by assuming that tokens entering from a neighbouring component are in some way observable

fault diagnosis

- Here: relax diagnosis goal!
- Enumerate only the minimal traces containing sequences of events that <u>must</u> have happened for OBS (or in distributed case OBS)
 to be allowable
- i.e. do not expand traces with transitions that do not lead to satisfaction of constraints necessary for occurrence of observbale event and only use minimal number of tokens

Centralised fault diagnosis

- Set E_{Min}(OBS) of minimal explanations allows us
 - -to decide if a fault happened for sure
 - but if fault t_{f1} is not included in any trace in $\mathcal{E}_{Min}(OBS)$ we do not know whether this is because plant is free of fault t_{f1} or whether fault t_{f1} has occurred but has not yet had any observable effect

Fault detectability

- Fault detection with minimal explanations is reasonable if we assume (as for most FDI papers)
 - that each fault leads to an observable effect after a bounded number of events (no unobservable cycles between fault and observable event), and each choice after a fault event leads to an observation
 - faults cannot be anticipated, i.e. input places of faults are choice places that also lead to a "good" Paris, October 14, 2005 PN based distributed diagnosis 37

Construction of minimal explanations

- assume $OBS = \{t_1^o\}$, 1st observation at time $q(t_1^o)$
- necessary constraint for execution of event t_1^o is marking by at least 1 token of each place p_{in}^k in $Pre(t_1^o)$
- this in turn requires that for each place p_{in}^k at least one of the input transitions (determined by Post(., p_{in}^k) of p_{in}^k has fired prior to q(t₁^o)

Construction of minimal explanations centralised agent

- recursive enumeration alternatingly using Pre and Post ends when backward path reaches a place that contains enough tokens according to the initial marking
- if there are unobservable cycles then it is possible that on a first visit to an intially marked place there are not enough tokens, but that repeating cycle several times may provide enough tokens

Construction of minimal explanations

- BUT: if no unobservable cycles with choice places exist, then number of tokens in cycle cannot unobservably change
- this ensures validity of stopping criterion: stop when enough tokens are available; remove trace form set of possible explanations if place is visited for 2nd time and still not enough tokens are found
- Each minimal explanation consists of a trace that GANNOT containstany observable event⁴⁰

Construction of minimal explanations

 remark: assumption no unobservable cycles with choice places, ensures that no problems due to tokens moving unobservably from one cycles containing several initially marked places

Construction of set of minimal explanations

- How large is set E_{Min}(OBS) of minimal explanations?
- At each backward choice place each possible input transition must be explored: size exponential in #backward choice places

Construction of set of minimal explanations

- if Petri net does not have unobservable cycles then one can prove that result is independent of order in which different explanations are explored
- Hence: result can be proven correct and complete (all obtained traces are minimal explanations and all minimal explanations are found)

Recursive calculation of set of minimal explanations

- After 1st observation $OBS = \{t_1^o\}$ calculate for each trace $\tau_{exp}^n \in \mathcal{E}_{Min}(OBS)$ the marking $m_1(\tau_{exp}^n)$ reached by executing τ_{exp}^n starting from m_0
- For each $\tau_{exp}{}^{n}$ use marking $m_1(\tau_{exp}{}^{n})$ as the new initial marking for calculating the set of minimal explanations when the 2nd observed event $t_2{}^{o}$ is detected

Recursive calculation of set of minimal explanations

- if starting from $m_1(\tau_{exp}{}^n)$ the trace $\tau_{exp,2}{}^m$ is a possible minimal explanation of $t_2{}^0$
- the concatenatation τ_{exp}^{n} . $\tau_{exp,2}^{m}$ is a minimal explanation of observation $OBS = \{t_1^{o}, t_2^{o}\}$

Recursive calculation of set of minimal explanations

- the set of all the minimal explanations for $\{t_1^{\,\,o},\,t_2^{\,\,0}\}$ is $E_{Min}(\{t_1^{\,\,o},\,t_2^{\,\,0}\}$)
- each trace $\tau_{exp}{}^n \in \mathcal{E}_{Min}(\{t_1{}^o, t_2{}^o\})$ leads to a new marking $m_2(\tau_{exp}{}^n)$ to be used as initial marking

when the 3rd observed event is detected

Construction of minimal explanations distributed case

- Problem in distributed case:
- local calculation: when to stop if backward search path leads to input place in $P_{\text{IN,i,i}}$?
- whenever backward search for possible minimal explanation $\tau_{exp}{}^{n}$ leads to input place $p_{in,k} \in P_{IN,i,j}$, stop backward search and add information to $\tau_{exp}{}^{n}$ that component j must have some local explanations that put at least one (additional) token in $p_{in,k}$

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Construction of minimal explanations distributed case

- When must token arrive in p_{in,k}?
- Prior to occurrence time of t_n⁰ that leads to generation of trace ending in p_{in,k}
- Implies assumption: all components have perfectly synchronised clocks, and can time stamp all observed event accurately

Global clock!

Some particularities of our model

- Unlike many other distributed analyses (Fabre, Jard, Su,...)
- we assume global clock available
- but each agent only knows local model and interactions with neighbouring models
- justification:
 - GPS timer sufficiently accurate for applications,
 - but many reconfigurations make it difficult for each agent to know global model

Goals of distributed fault detection

- From time to time local agnets should exchange enough information
- so that local diagnosis result in component i detects all the local faults that global diagnoser would detect at same time
- i.e. after communication between agents local diagnosis = projection of global diagnosis

Questions to be answered

- when to communicate?
- what is minimal information to be exchanged between agents?
- what assumptions must be made on model and on protocol to make this possible?

Outline

- Petri net model
- model-based diagnosis:
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 - problem formulation in distributed case
 - minimal explanations
- distributed diagnosis algorithm

Distributed diagnosis

- arbitrarily select time q > 0
- for each agent i
 - sense $OBS_i(q)$ such that all observations in $OBS_i(q)$ occur prior to q, are local to component i, and all observable events in i prior to q are included
 - enumerate locally in agent i the set of local minimal explanations $\mathcal{E}_{Min}(OBS_{i}(q))$
- at time q "stop the clock" and exchange information between all agents until stopping criterion

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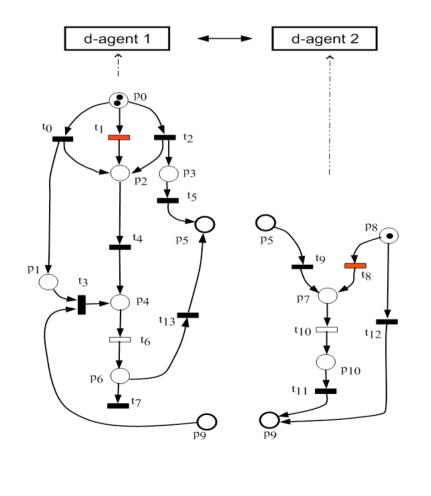
Distributed diagnosis

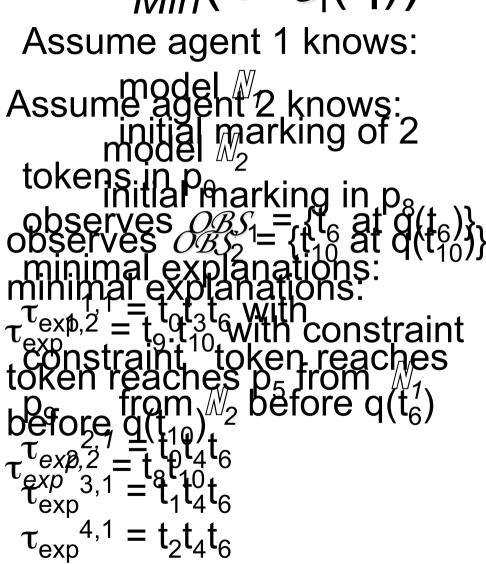
"stop the clock" means assumption of no communication delays infinitely fast processor Moreover we assume that the global clock is perfect, i.e. all events over all components can be perfectly ordered after exchanging information between agents

choice of q

- since there are no common events between agents, q must be treated as an arbittrary time
- it can be triggered by any event in any component, and will then be accepted by all other agents
- or it may be set by an outside supervisor

local calculation of $\mathcal{E}_{Min}(OBS_{i}(q))$

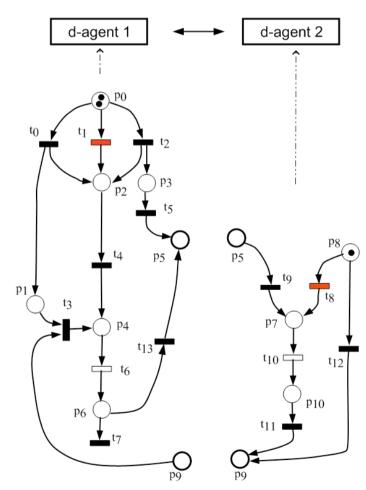




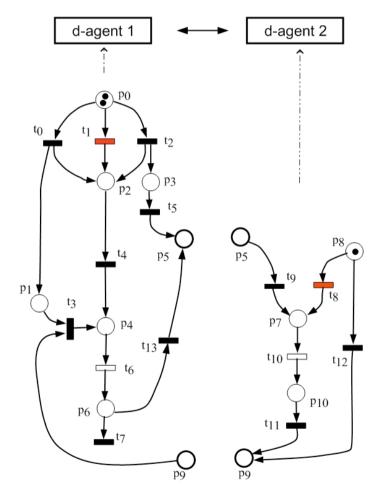
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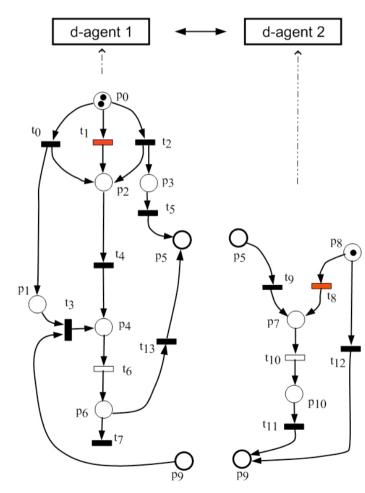
d-agent 2 d-agent 1 if $q(t_{10}) < q$ and agent 2 has not sensed any other event before q, then agent 2 can initiate information exchange with agent 1: minimal explanation $\tau_{exp}^{1,2} = t_9 \cdot t_{10}$ p5 p5 asks agent 1 which of its p₁ t3 minimal explanations can put **p**7 token in p_5 before $q(t_{10})$ t10 [t12 agent 1 calculates that $\tau_{exp}^{4,1}$ can p10 p6 do this whatever the value of $q(t_6)$



Hence τ $e_{xp}^{1,2} = t_9 \cdot t_{10}$ and $\tau_{exp}^{4,1} = t_2 t_4 t_6$ together form a global minimal explanation without any fault event $\tau_{exp}^{2,2}$ can be combined with $\tau_{exp}^{2,1}$ forming a global explanation containing fault t_8 $\tau_{exp}^{2,2}$ can be combined with $\tau_{exp}^{3,1}$ forming a global explanation containing both faults t_1 and t_8



Moreover agent 1 will respond to the question of agent 2 that it can always put a token into place p₅ via the unobservable trace $t_2 t_5$ This requires agent 1 to do a backward calculation finding a minimal explanation for a token in p_5



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In order to form the complete set of global minimal explanations agent 1 must also ask agent 2 if it can put a token into p₉ before time $q(t_6)$ in order to allow $\tau_{exp}^{1,1}$ Agent 2 will respond that this can via the unobservable trace t_{12} (excluding $\tau_{exp}^{1,2}$ but not excluding $\tau_{exp}^{2,2}$), and that moreover it can also be combined with $\tau_{exp}^{1,2}$ provided $q(t_{10}) < q(t_6) (< q)$ PN based distributed diagnosis 60

- In previous example question-and-answer session comes to conclusion as soon as all pairs of local minimal epxlanations have been validated or rejected
- This depends on assumption that no token can pass unobserved from input place to output place (e.g. observable transition t₁₀ on path from p₅ to p₉)

 assumption of at least one observable transition on any path from input to output place in any component can be dropped at price of longer calculations:

need to use upper bounds on number of tokens in output places to guarantee that iterations eventually stop

- More than 2 components:
- communication protocol must ensure "fairness" i.e. each agent can only exchange a bounded number of messages with one of its neighbours before starting a round of communication exchange with each of its other neighbours.
- This guarantees that algorithm eventually decides for each possible combination of local explanations whether they are global minimal explanations

 this "fairness" requirement is analogous to requirements in asynchronous solution of lsets of algebraic equations

(like asynchronous Gauss-Seidel)

Recursive distributed fault detection

- after completion of communication round at time q
- each local agent can calculate the marking reached after executing the local projection of each of the accepted global minimal explanations
- these marking can be used a initial marking for the next run of the distirbuted fault detection algorithm

Outline

- Petri net model
- model-based diagnosis:
 - problem formulation in centralised case
 - problem formulation in distributed case
 - minimal explanations
- distributed diagnosis algorithm
- conclusion

conclusion

- distributed fault diagnosis can be implemented using local agents
 - knowing only local model and local initial marking
 - sensing local event occurrences (incl. times)
 - calculating recursively local minimal epxlanations
 up to some time q when communication between all agents is initiated by supervisory protocol
 this allows each local agent to decide which local minimal epxlanations are globally valid

extensions

 include event timing information in model (e.g. using time Petri net models for some components) in order to further reduce set of minimal explanations

extensions

- model based fault detection can only detect faults that are explicitly modelled, but often the number of potential faults is so large that this is computationally infeasible
- probabilistic models where most likely models are checked first become attractive

extensions

 both timed models and probabilistic models require combination of forward analysis (using unfoldings and configurations) and backward analysis

