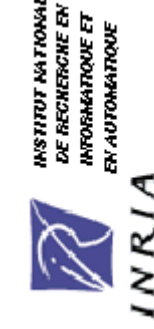


Methods and algorithms for the performance analysis of large systems

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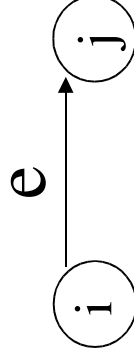
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Motivations

- Continuous Time Markov Chains (CTMC): performance analysis of dynamic systems.
- Huge CTMCs: compact and structured model. **State space explosion.**
- Computing stationary and transient solutions: **sophisticated algorithms and data structure are needed.**

SAN

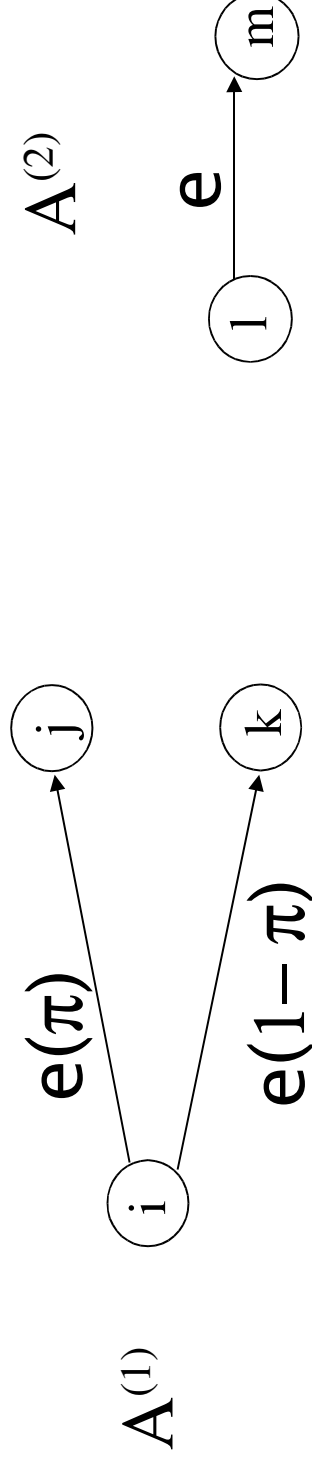
- Change of the global state (combination of the local states of each automata):
 - | local or synchronising event, transition rate λ .
- Local event:
 - | Change of the local state of only one automaton.



- Synchronising event:
 - | Simultaneous change of the local state of more than one automaton.

SAN

- Probability of occurrence when the occurrence of the same event from a given departure state can lead to different arrival states.



SAN



- Other way of interacting: **functional rates**.
- The transition rate of an event can be a function of the global state of the SAN.
- Can apply to local or synchronising events.

SAN descriptor

- SAN: modular description and manipulation.
Generator of the Markov chain = **descriptor Q**.

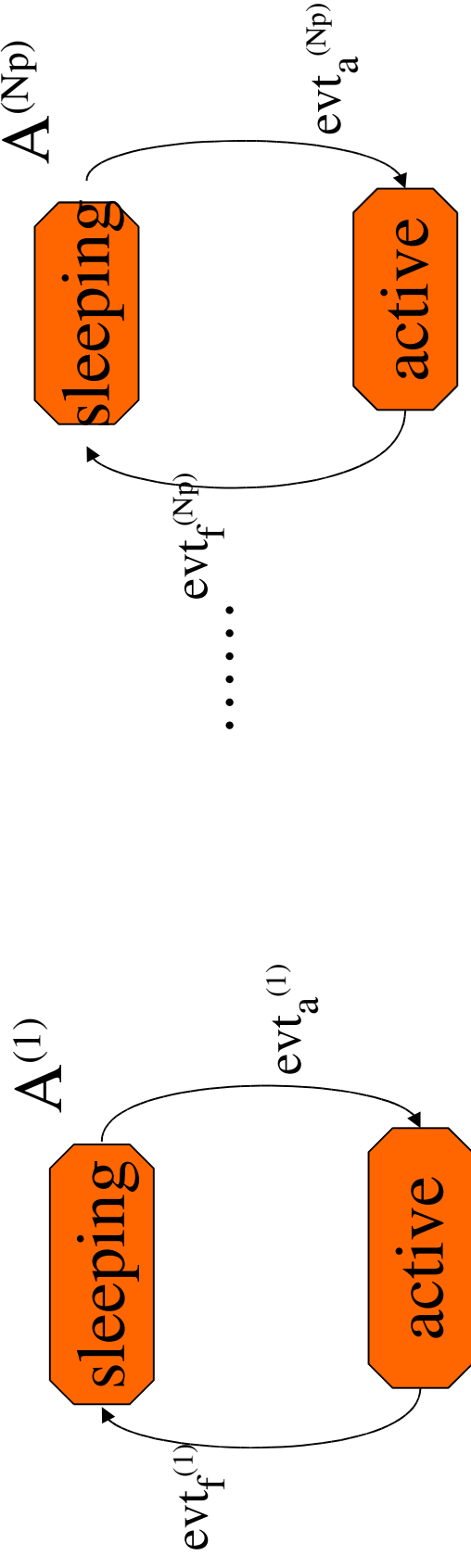
$$Q = D + \sum_{i=1}^N \sum_{e=1}^E Q_e^{(i)} + \sum_{j=1}^{N+2E} \sum_{i=1}^N Q_j^{(i)}$$

- Tensor sum: local part.
- Tensor product: synchronising part - E sync events.
- Use of generalised tensor algebra (functional matrices) [Plateau, Fernandes ...].

SAN

study example - mutex1

- N_p processes, N_r resources - model with functions



For $i=1..N_p$, $evt_a^{(i)}$ is a local event « acquiring a resource », rate $\lambda_i f(A^{(1)}, \dots, A^{(N_p)})$.

$f(A^{(1)}, \dots, A^{(N_p)})$ is a function of the number of automata in state active.

If this number is greater than or equal to N_r , f equals 0, otherwise 1.

For $i=1..N_p$, $evt_f^{(i)}$ is a local event « freeing a resource », rate μ_i .

SAN

example - mutex1 - descriptor

- For $i=1..Np$, define $Q_l^{(i)} = \begin{pmatrix} - & i f & i f \\ \mu_i & & -\mu_i \end{pmatrix}$
- Then the **descriptor** (algebraic formulation of the transition matrix) is:

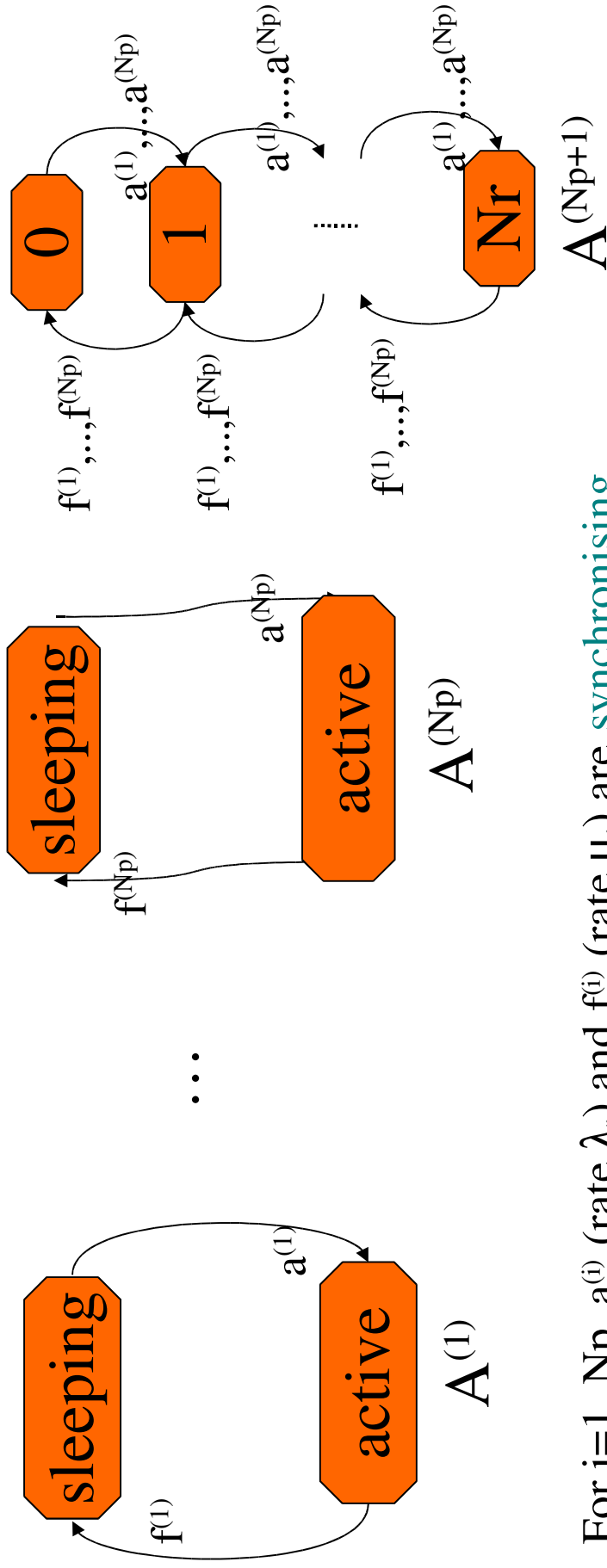
$$Q = Q_l^{(1)} \oplus_g Q_l^{(2)} \oplus_g \dots \oplus_g Q_l^{(Np)}$$

- If the N automata were independent (i.e. the function f constant), the descriptor would be the same, with classical tensor sum.

SAN

example - mutex2

- N_p processes, N_r resources - synchronising events



For $i=1..N_p$, $a^{(i)}$ (rate λ_i) and $f^{(i)}$ (rate μ_i) are **synchronising events**, implying automata $A^{(i)}$ and $A^{(N_p+i)}$.

SAN

comparison mutex1/mutex2

- Product state space:

$$|S_1| = 2^{Np} \text{ and } |S_2| = 2^{Np} * (Np + 1)$$

- Reachable state space size: $|S_1|$ and $|S_2|$

equal to the number of solutions of

$$\sum_{i=1}^{Np} x^{(i)} \leq Nr, \text{ with } x^{(i)} = 0 \text{ or } 1$$

- Descriptor: Mutex1 - (Np-1) tensor sums.
- Mutex2 - $2(Np)^2$ tensor products of very sparse matrices.

On-going work



... and items of interest

Modelling large computer systems formalisms

- Queuing networks
- Petri nets
- Process algebra
- Automata networks
- others:
 - UML
 - Mobius
 - combinations of PA and PN, PA and SAN?

Modelling large computer systems storing S

- **Reachability function**
 - Access to the reachable states in constant time.
 - Example: PEPS.
- **Multi-level approaches**
 - Decision diagrams [Ciardo, Miner].
 - Efficient generation and storage of S .

Modelling large computer systems

lumping of models

reduce the complexity of the Markov chain, using **replications of components** and **exact lumping**.

- Aggregation from state space analysis
- Aggregation from the high level formalism:
 - | [Siegle], [Buchholz]
 - | [Benoit]: SANs with replicas and tensor structure of the aggregate
 - | extension and combination

Modelling large computer systems

numerical algorithms

- Context: iterative methods
- tensor: work on a product state space \hat{s}
- Probability vectors \hat{s} « extended » (the size of \hat{s})
- Q: descriptor of the SAN
$$Q = \sum_{j=1}^{N+2E} \begin{bmatrix} N \\ Q_j^{(i)} \\ i=1 \end{bmatrix}$$
- sparse implementation, in parallel?

PEPS to developp

- All these algorithms are implemented in **PEPS** (Performance Evaluation of Parallel Systems)
 - SAN editor
 - Describes easily SANs with replicas, and replicated states with identical behavior (queuing networks).
 - Computes and tests the descriptor.
 - Implements the basic multiplications, with optimizations for the function evaluation (automata reordering, ...).
 - Various iterative methods (power, Jacobi, GMRES).
 - Allows to output sparse format (HBF).
 - Compute performance indices.

Other Perspectives



- **Graphical interface** in Java for PEPs: more conviviality, easy to use [Mathieu Le Coz]
- **Discrete-time SANs**: we are working on a new formalism; develop new numerical methods.
- **Compute bounds**
- **Simulation ?**

Annexe (1)

tensor product

$$\text{Product } A \otimes B = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \otimes \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}b_{11} & a_{11}b_{12} & a_{11}b_{13} & a_{12}b_{11} & a_{12}b_{12} & a_{12}b_{13} \\ a_{11}b_{21} & a_{11}b_{22} & a_{11}b_{23} & a_{12}b_{21} & a_{12}b_{22} & a_{12}b_{23} \\ a_{11}b_{31} & a_{11}b_{32} & a_{11}b_{33} & a_{12}b_{31} & a_{12}b_{32} & a_{12}b_{33} \\ a_{21}b_{11} & a_{21}b_{12} & a_{21}b_{13} & a_{22}b_{11} & a_{22}b_{12} & a_{22}b_{13} \\ a_{21}b_{21} & a_{21}b_{22} & a_{21}b_{23} & a_{22}b_{21} & a_{22}b_{22} & a_{22}b_{23} \\ a_{21}b_{31} & a_{21}b_{32} & a_{21}b_{33} & a_{22}b_{31} & a_{22}b_{32} & a_{22}b_{33} \end{bmatrix}$$

Annexe (2)

tensor sum

$$\text{Sum } A \oplus B = A \otimes \text{Idn}_B + \text{Idn}_A \otimes B$$

$$= \begin{pmatrix} a_{11} + b_{11} & b_{12} & b_{13} & a_{12} & \\ b_{21} a_{11} + b_{22} & & b_{23} & a_{12} & \\ b_{31} & b_{32} a_{11} + b_{33} & & & a_{12} \\ a_{21} & & & a_{22} + b_{11} & b_{12} & b_{13} \\ & a_{21} & & b_{21} a_{22} + b_{22} & b_{23} & \\ & & & b_{31} & b_{32} a_{22} + b_{33} & \end{pmatrix}$$

Annexe (3)

generalised tensor product

$$\begin{array}{c}
 \left[\begin{array}{cc} a_{11}(k) & a_{12}(k) \\ a_{21}(k) & a_{22}(k) \end{array} \right] \otimes_{\mathbb{R}} \left[\begin{array}{ccc} b_{11}(h) & b_{12}(h) & b_{13}(h) \\ b_{21}(h) & b_{22}(h) & b_{23}(h) \\ b_{31}(h) & b_{32}(h) & b_{33}(h) \end{array} \right] \\
 \hline
 \left(\begin{array}{ccc|ccc|ccc}
 a_{11}(1)b_{11}(1)a_{11}(1) & b_{12}(1)a_{11}(1) & b_{13}(1) & a_{12}(1)b_{11}(1)a_{12}(1) & b_{12}(1)a_{12}(1) & b_{13}(1) & a_{12}(2)b_{21}(1)a_{12}(2) & b_{22}(1)a_{12}(2) & b_{23}(1) \\
 a_{11}(2)(1) & a_{11}(2) & a_{11}(2) & a_{12}(3)b_{31}(1)a_{12}(3) & b_{32}(1)a_{12}(3) & b_{33}(1) & a_{21}(1)b_{11}(2)a_{21}(1) & b_{12}(2)a_{21}(1) & b_{13}(2) \\
 a_{21}(2) & a_{21}(2) & a_{21}(2) & a_{22}(2) & a_{22}(2) & a_{22}(2) & a_{22}(2) & a_{22}(2) & a_{22}(2) \\
 a_{21}(3) & a_{21}(3) & a_{21}(3) & a_{22}(3) & a_{22}(3) & a_{22}(3) & a_{22}(3) & a_{22}(3) & a_{22}(3)
 \end{array} \right)
 \end{array}$$