

TAPAS: a Tool for Stochastic Evaluation of Large Interdependent Composed Models with Absorbing States

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Agenda

- Context & objectives
- Addressed models and measures
- Input format: (a restriction of) Stochastic Automata Network
- Descriptor matrices and vectors through Tensor Trains
- Demo
- Next steps

Context & Objectives

Context:

- Reliability CTMC models (with focus on limiting behavior) where there are **absorbing states**, e.g., system failure states
- **Large models**, in particular: the system model comprises several synchronized submodels, each with a relatively small state-space

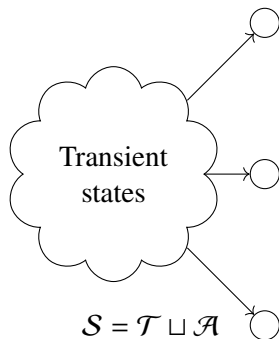
How our tool TAPAS (Tool for Stochastic Evaluation of Large Interdependent Composed Models with Absorbing States) contributes:

- it **evaluates** performability measures
- it exploits (compressed) **implicit representation** of all the matrices and vectors

Addressed models and measures

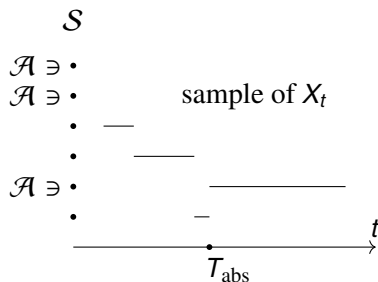
Models & measures of interest

- Consider the CTMC $\{X_t\}_{t \geq 0}$ with absorbing states \mathcal{A} , where $\forall t. X_t \in \mathcal{S}$ and $X_0 \notin \mathcal{A}$



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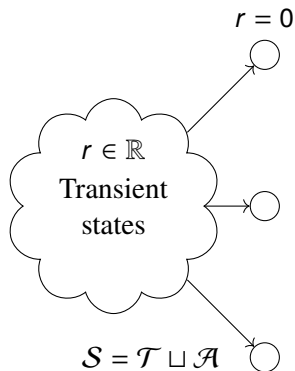
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Infinitesimal Generator Matrix

$$\begin{bmatrix} Q_{\mathcal{T}} & v_1 & v_2 & \dots & v_{|\mathcal{A}|} \\ \hline 0 & \dots & 0 & 0 & 0 & 0 & 0 \\ \vdots & & \vdots & \vdots & \vdots & \vdots & \vdots \\ \hline 0 & \dots & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

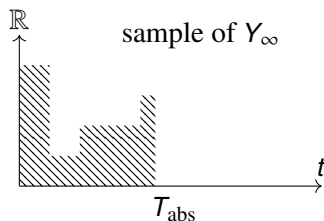
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- Define, for $i \in \mathcal{S}$, a reward vector $r = [r_i]$ such that $r_i = 0$ if $i \in \mathcal{A}$



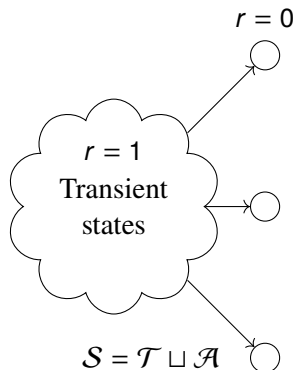
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- Define $Y_\infty := \int_0^\infty r_{X_t} dt$



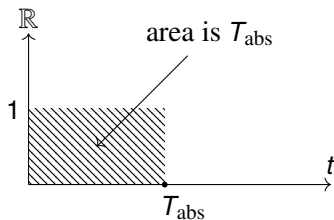
Models & measures of interest

- The moments of Y_∞ ,
i.e., $\mathcal{M}_k := E[Y_\infty^k]$,
e.g., $\text{MTTA} = E[T_{\text{abs}}] = E[Y_\infty]$
where $r_i = 1$ for $i \in \mathcal{T}$



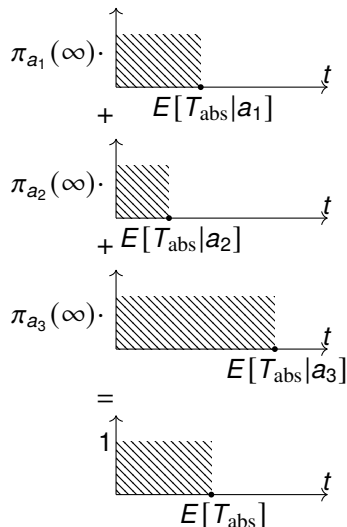
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where $r_i = 1$ for $i \in \mathcal{T}$
- $\pi_{\mathcal{B}}(\infty)$, i.e., the probability that X
is absorbed in $\mathcal{B} \subset \mathcal{A}$
- $E[Y_\infty | \mathcal{B}]$ for $\mathcal{B} \subset \mathcal{A}$, e.g., for
 $\{a_1\} \subseteq \{a_1, a_2, a_3\}$ evaluate
 $E[T_{\text{abs}} | \{a_1\}]$



Measures evaluation for explicit models

TAPAS works with the following formalization:

measure	evaluation
\mathcal{M}_k	$\begin{cases} (Q - S)x^{(1)} = r, \\ (Q - S)x^{(i)} = \text{diag}(r)x^{(i-1)}, \text{ for } i = 2, \dots, k, \end{cases}$ then $\mathcal{M}_k = k!(-1)^k \pi(0) \cdot x^{(k)}$
$\pi_{\mathcal{B}}(\infty)$	$(Q - S)x^{(1)} = Qe_{\mathcal{B}}$, then $\pi_{\mathcal{B}}(\infty) = -\pi(0) \cdot x^{(1)}$,
$\text{MRTA}_{ \mathcal{B}}$	$\begin{cases} (Q - S)x^{(1)} = Qe_{\mathcal{B}}, \\ (Q - S)x^{(2)} = \text{diag}(r)x^{(1)}, \end{cases}$ then $\text{MRTA}_{ \mathcal{B}} = (\pi(0) \cdot x^{(2)}) / (-\pi(0) \cdot x^{(1)})$

where the shift matrix S is defined so that, being $r_a = 0$ for $a \in \mathcal{A}$,

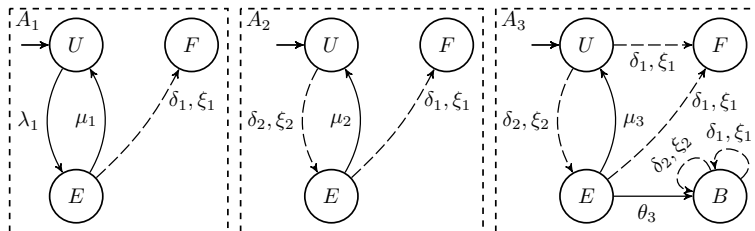
$$\pi(0) \cdot (Q - S)^{-1} \cdot r = \pi_{\mathcal{T}}(0) \cdot Q_{\mathcal{T}}^{-1} \cdot r_{\mathcal{T}}$$

there is no need to distinguish transient from absorbing states in the labeling

Stochastic Automata Network (SAN)

SAN models

A_1, \dots, A_n define the stochastic process $\tilde{X} = (X^{(1)}, \dots, X^{(n)})$, that is indistinguishable from X , where a synchronization transition (dashed arrow) is enabled if and only if it is enabled in all the automata, e.g., ξ_2 can “fire” if and only if A_2 is in U and (A_3 is in U or in B)



The state-space (small) exploration of each automaton is performed independently from the others

To evaluate the measures of interest it is required the existence of a path from each state to the absorbing states

Addressed technical challenge

All the quantities involved in the computations (\tilde{Q} , \tilde{S} , $\tilde{x}^{(i)}$, $\tilde{\pi}(0)$, \tilde{r}) can be expressed as a sum of Kronecker products (see demo) but

the issue is that each matrix-vector multiplication squares the number of addends in $\tilde{x}^{(i)}$. Thus, there is an exponential growth of memory consumption in iterative linear system solvers.

Tensor Trains (TT)

Proposed solution

- Compressed matrices and vectors were already investigated in the literature for CTMC performance and availability models (no absorbing states)
- Novelty of TAPAS:** TT-based representation for CTMC reliability models (with absorbing states)

$$\mathcal{A}_{i_1 \dots i_d} = \underbrace{G_1[i_1]}_{1 \times r} \underbrace{G_2[i_2]}_{r \times r} \dots \underbrace{G_d[i_d]}_{r \times 1}$$

An example of computing one element of 4-dimensional tensor:

$$\mathcal{A}_{2423} = \begin{matrix} G_1 \\ \text{row } i_1=2 \end{matrix} \times \begin{matrix} G_2 \\ \text{row } i_2=4 \end{matrix} \times \begin{matrix} G_3 \\ \text{row } i_3=2 \end{matrix} \times \begin{matrix} G_4 \\ \text{row } i_4=3 \end{matrix}$$

- Standard iterative methods to evaluate $\tilde{\chi}^{(i)}$ fail because in

$$\tilde{\chi}^{(i,j+1)} = \tilde{\chi}^{(i,j)} + \Delta \tilde{\chi}^{(i,j)}$$

the TT-ranks can grow too quickly

Proposed solution

- Thus, **ad hoc solution methods** have been investigated (6 so far):

Method	Transposed	Published	TT	Exponential Sums
tt-regular-splitting			✓	✓
amen(t)	✓	✓	✓	✓
gmres(t)	✓		✓	
tt-expsumst	✓		✓	✓

- Each $\tilde{\chi}^{(i,j+1)}$ update has to be re-compressed (exploiting TT round, that is based on SVD), and the TAPAS user can set the corresponding tolerance ttol. This parameter impacts mainly on memory occupancy
- The result tolerance, i.e., tol, can also be set. This parameter impacts mainly on time and accuracy

Demo

Next steps

- Exploit the tool to evaluate additional performability measures of interest, with focus on specific application domains
- Fully adhere to the Stochastic Automata Network (SAN) formalism, resorting to generalized Kronecker algebra theory
- Develop **new features**, e.g. to allow steady-state analysis, so to emphaddress availability-related measures, and to import the model description, as elaborated by other tools
- Integration of TAPAS in other tools, in addition to MATLAB, possibly open source ones, to promote wider usability

Thank you
Questions?