

An Efficient Lower Transition Rate Operator for the Sensitivity Analysis of Continuous-Time Markov Chains*

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Continuous-time Markov chains (CTMCs) are popular probabilistic models that describe the uncertain evolution of a dynamical system whose time domain is the non-negative reals $\mathbb{R}_{\geq 0}$. We assume that this system takes values in a non-empty and finite *state space* \mathcal{X} , with size $n := |\mathcal{X}|$. When a CTMC is *time-homogeneous*, it can be characterised—up to its initial distribution—by a single *rate matrix* Q . This is a $n \times n$ matrix with entries $Q(x, y) \in \mathbb{R}$, such that for all $x \in \mathcal{X}$, $\sum_{y \in \mathcal{X}} Q(x, y) = 0$ and $Q(x, y) \geq 0$ for all $y \in \mathcal{X}$ with $x \neq y$. We use \mathcal{R} to denote the set of all rate matrices.

When faced with uncertainty about the specification of a CTMC—be it with respect to the numerical parameters, or structural assertions like Markovianity or time-homogeneity—a sensitivity analysis may be performed using a continuous-time *imprecise*-Markov chain (CTIMC). This model can be seen as a *set* $\mathbb{P}_{\mathcal{Q}}$ of stochastic processes that are *consistent* with a non-empty bounded *set* $\mathcal{Q} \subset \mathcal{R}$ of rate matrices. One may consider different types of CTIMCs by including different types of models in $\mathbb{P}_{\mathcal{Q}}$. For example, one may include all time-homogeneous CTMCs characterised by the elements $Q \in \mathcal{Q}$. However, one could additionally include non-homogeneous CTMCs, or even more general—non-Markovian—processes; see Krak [1] for an analysis of these different types. Given a CTIMC $\mathbb{P}_{\mathcal{Q}}$, the associated (conditional) lower- and upper expectations $\underline{\mathbb{E}}_{\mathcal{Q}}[\cdot | \cdot]$ and $\overline{\mathbb{E}}_{\mathcal{Q}}[\cdot | \cdot]$ provide tight lower- and upper bounds, for any inference of interest, with respect to every model included in $\mathbb{P}_{\mathcal{Q}}$. To make this practicable, we need efficient methods to compute these lower- and upper expectations; by the well-known *conjugacy* property $\overline{\mathbb{E}}_{\mathcal{Q}}[\cdot | \cdot] = -\underline{\mathbb{E}}_{\mathcal{Q}}[-\cdot | \cdot]$, it suffices to focus on only the lower expectations.

Following Škulj [2], we associate with \mathcal{Q} its corresponding *lower transition rate operator* \underline{Q} , which is a map $\underline{Q} : f \mapsto \underline{Q}f$ from $\mathbb{R}^{\mathcal{X}}$ to $\mathbb{R}^{\mathcal{X}}$, where for all $f \in \mathbb{R}^{\mathcal{X}}$ and $x \in \mathcal{X}$, we let $\underline{Q}f(x) := \inf_{Q \in \mathcal{Q}} Qf(x)$. Then for any $f \in \mathbb{R}^{\mathcal{X}}$ and $t \in \mathbb{R}_{\geq 0}$, the quantity $\underline{\mathbb{E}}_{\mathcal{Q}}[f(X_t) | X_0]$ can be conservatively bounded by the solution $\underline{f}_t \in \mathbb{R}^{\mathcal{X}}$ to the initial value problem

$$\frac{d\underline{f}_t}{dt} = \underline{Q}\underline{f}_t \quad \text{for all } t \in \mathbb{R}_{\geq 0}, \text{ with } \underline{f}_0 = f. \quad (1)$$

That is, it holds that $\underline{f}_t(x) \leq \underline{\mathbb{E}}_{\mathcal{Q}}[f(X_t) | X_0 = x]$ for all $t \in \mathbb{R}_{\geq 0}$ and $x \in \mathcal{X}$; crucially however, this bound is satisfied with equality under some relatively mild assumptions on $\mathbb{P}_{\mathcal{Q}}$ and \mathcal{Q} [2, 1]. Hence in those cases, solving (1) amounts to computing the lower expectation of interest exactly. The differential equation (1) may be solved e.g. by iterative methods described by Škulj [2], Krak [1]. To apply these methods, we need an (ideally efficient) way to evaluate $\underline{Q}f$ for any $f \in \mathbb{R}^{\mathcal{X}}$.

We here consider a special case that we feel is important for sensitivity analysis, where we construct a simple perturbation model for a given rate matrix Q ; we let $\mathcal{Q} := \mathcal{B}(Q, r) := \{Q' \in \mathcal{R} : \|Q - Q'\|_{\infty} \leq r\}$ be the closed ball in \mathcal{R} of radius $r \in \mathbb{R}_{\geq 0}$ around Q . This \mathcal{Q} satisfies the required properties to make the solution of (1) coincide with the lower expectation of interest [1, Prop. 6.20 and Cor. 6.25]. Our main current contribution is Algorithm 1, which enables the efficient computation of $\underline{Q}f$ in this case; for a proof of correctness, see Krak [1, Prop. 6.21]. It is easily seen that the runtime complexity of this algorithm is $O(n^2)$. This enables the exceptionally efficient computation of exact bounds on inferences under this perturbation model.

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input: Rate matrix  $Q \in \mathcal{R}$ , radius  $r \in \mathbb{R}_{\geq 0}$ , function  $f \in \mathbb{R}^{\mathcal{X}}$ 
output: Function  $\underline{Q}f \in \mathbb{R}^{\mathcal{X}}$ , where  $\underline{Q}$  corresponds to  $\mathcal{Q} = \mathcal{B}(Q, r) \subset \mathcal{R}$ 
 $(x_1, \dots, x_n) := \text{SortDecreasing}(\mathcal{X}, f)$ ; // sort  $\mathcal{X}$  such that  $f(x_i) \geq f(x_{i+1})$  for all  $i \in \{1, \dots, n-1\}$ 
for  $x \in \mathcal{X}$  do
   $r_0 := r/2$ 
  for  $i \in \{1, \dots, n\}$  do
    if  $x_i = x$  then  $\Delta_i := r_{i-1}$  else  $\Delta_i := \min\{r_{i-1}, Q(x, x_i)\}$ 
     $r_i := r_{i-1} - \Delta_i$ 
   $\underline{Q}f(x) := Qf(x) - \sum_{i=1}^n \Delta_i (f(x_i) - f(x_n))$ 
return  $\underline{Q}f$ 

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Algorithm 1: Efficiently evaluate $\underline{Q}f$.

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References

- [1] Thomas Krak. *Continuous-Time Imprecise-Markov Chains: Theory and Algorithms*. PhD thesis, 2021.
- [2] Damjan Škulj. Efficient computation of the bounds of continuous time imprecise Markov chains. *Appl. Math. Comput.*, 250:165–180, 2015.

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