

Model Checking Continuous Time Markov Chains

Adnan Aziz

The University of Texas at Austin

and

Kumud Sanwal

Lucent Technologies

and

Vigyan Singhal

Tempus-Fugit, Inc.

and

Robert Brayton

The University of California at Berkeley

We present a logical formalism for expressing properties of continuous time Markov chains. The semantics for such properties arise as a natural extension of previous work on discrete time Markov chains to continuous time. The major result is that the verification problem is decidable; this is shown using results in algebraic and transcendental number theory.

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1. INTRODUCTION

Recent work on formal verification has addressed systems with stochastic dynamics. Certain models for discrete time Markov chains have been investigated in [Hansson and Jonsson 1994; Courcoubetis and Yannakakis 1988]. However, a large class of stochastic systems operate in continuous time. In a generalized decision and control

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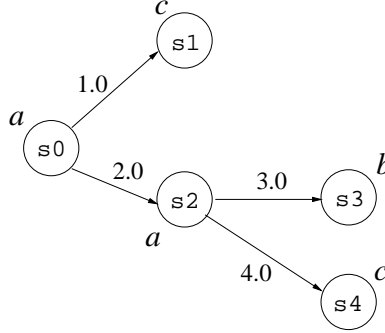


Fig. 1. A continuous time Markov chain: $S = \{s_0, s_1, s_2, s_3\}$, $A = \{a, b, c\}$. Only edges with positive weights are shown.

framework, continuous time Markov chains form a useful extension [Ross 1983]. We propose a logic for specifying properties of such systems, and describe a decision procedure for the model checking problem. Our result differs from past work on verifying continuous time Markov chains [Alur et al. 1991] in that quantitative bounds on the probability of events can be expressed in the logic.

2. CONTINUOUS TIME MARKOV CHAINS

Formally, a continuous time Markov chain M is a 4-tuple (S, Λ, A, θ) , where $S = \{s_1, s_2, \dots, s_n\}$ is a finite set of *states*, Λ is the *transition rate matrix*, A is a finite set of *outputs*, and $\theta : S \mapsto A$ is the *output function*. The transition rate matrix Λ is an $|S| \times |S|$ matrix. The off-diagonal entries are nonnegative rationals; the diagonal element $\lambda_{j,j}$ is constrained to be $-(\sum_{i \neq j} \lambda_{j,i})$. Consequently, the row sums of Λ are zero.

An example of a continuous time Markov chain is presented in Figure 1.

The notion of a continuous time Markov chain can be generalized to include rate matrices whose entries vary with time. We will restrict our attention to *homogeneous* continuous time Markov chains, for which the rate matrices are constants.

At state s_j , the probability of making a transition to state s_k (where $k \neq j$) in time dt is given by $\lambda_{j,k} dt$. This is the basis for formulating a stochastic differential equation for the evolution of the probability distribution whose solution is

$$D(t) = e^{\Lambda^T t} \cdot D_0$$

Here D_0 is a column vector of dimension $|S|$, with the constraint that $\sum_i D_0[i] = 1$.

A *path* through M is a function on domain $[0, \infty)$ and range S . any state s , we will denote by U^s the set of all paths beginning at s . We will later see how to compute the probability of a set of paths Π starting from a given state s using the rate transition matrix; this probability will be denoted by $\mu^s(\Pi)$. Given a set β of functions from $[0, \infty)$ to A , we will abuse notation and refer to $\mu^s(\beta)$ when we mean the probability of the set of paths starting at s whose images under θ map to elements in β . We will not dwell on the technicalities of measure theory; all sets of paths considered in this paper will be measurable.

3. CSL SYNTAX AND SEMANTICS

Let $M = (S, \Lambda, A, \theta)$ be a continuous time Markov chain. In this section, we develop formal syntax and semantics for CSL (Continuous Stochastic Logic). This logic is inspired by the logic CTL [Emerson 1990], and its extensions to discrete time stochastic systems (pCTL [Hansson and Jonsson 1994]), and continuous time nonstochastic systems (tCTL [Alur et al. 1990]).

There are two types of formulae in CSL: state formulae (which are true or false in a specific state), and path formulae (which are true or false along a specific path). A state formula is given by the following syntax:

- (1) \mathbf{a} for $a \in A$
- (2) If f_1 and f_2 are state formula, then so are $\neg f_1, f_1 \vee f_2$
- (3) If g is a path formula, then $Pr_{>c}(g)$ is a state formula, where c is a rational between 0 and 1 expressed as the ratio of two binary coded integers.

Path formulas are formulas of the form

$\neg f_1 U_{[a_1, b_1]} f_2 U_{[a_2, b_2]} \cdots f_n$, where f_1, f_2, \dots, f_n are state formulas, and $a_1, b_1, \dots, a_{n-1}, b_{n-1}$ are nonnegative rationals expressed as the ratio of two binary coded integers.

CSL is the set of state formulae that are generated by the above rules.

Let f be a state formula, and g be a path formula. We now define the satisfaction relation (\models_M) using induction on the length of the formula. For a state formula f we use $\llbracket f \rrbracket_M$ to denote the set of states satisfying f .

- (1) f is of the form \mathbf{a} : $s \models_M f$ iff $\theta(s) = a$.
- (2) f is of the form $(\neg f_1)$: $s \models_M f$ iff $s \not\models_M f_1$.
- (3) f is of the form $(f_1 \vee f_2)$: $s \models_M f$ iff $s \models_M f_1$ or $s \models_M f_2$.
- (4) f is of the form $Pr_{>c}(g)$: $s \models_M f$ iff $\mu^s(\{\pi \in U^s \mid \pi \models_M g\}) > c$.
- (5) g is a path formula of the form $f_1 U_{[a_1, b_1]} f_2 U_{[a_2, b_2]} \cdots f_n$: $\pi \models_M g$ iff there exist real numbers t_1, \dots, t_{n-1} such that for each integer in $[1, n]$ we have $(a_i \leq t_i \leq b_i) \wedge (\forall t' \in [t_{i-1}, t_i]) (\pi(t) \in \llbracket f_i \rrbracket_M)$, where t_{-1} is defined to be 0 for notational convenience.

EXAMPLE 1. The formula $\phi = Pr_{>0.3}(aU_{[0.0, 4.0]}b)$ is a state formula for the Markov chain in Figure 1. It formally expresses the property that with probability greater than 0.3, the system will remain in a state where the output is a before making a transition before 4.0 time units have elapsed to a state where the output is b .

The probability of the set of paths starting at s_0 on which the output is a before becoming b before time 4.0 is given by the following integral:

$$\int_0^{4.0} e^{-3x} \cdot 3 \cdot \frac{2}{3} \cdot (1 - e^{-7 \cdot (4-x)}) \cdot \frac{3}{7} \cdot dx$$

This simplifies to $\frac{1}{14}(4 - 7 \cdot e^{-12} + 3 \cdot e^{-28})$, which is smaller than 0.3, and so ϕ is false at s_0 .

4. CSL MODEL CHECKING

The CSL model checking problem is as follows: given a continuous time Markov chain M , a state s in the chain, and a CSL formula f , is it the case that $s \models_M f$? In this section we establish that there is an effective procedure for model checking CSL.

THEOREM 1. *CSL model checking is decidable.*

PROOF. The nontrivial step in model checking is to model check formula of the form $Pr_{>c}(g)$. In order to do this we need to be able to effectively reason about the quantity $\mu^s(\{\text{paths } \pi \mid \pi(0) = s_0 \wedge \pi \models_M g\})$.

First we review some elementary algebra. An *algebraic complex number* is any complex number which is the root of a polynomial with rational coefficients. We will denote the set of algebraic complex numbers by \mathcal{A} . Properties of the algebraic numbers are derived in [Niven 1956]; of particular interest to us is the fact that they constitute a field, and that the real and imaginary parts of an algebraic number are also algebraic.

We will denote the set of complex numbers which are finite sums of the form $\sum_j \eta_j e^{\delta_j}$ where the η_j and δ_j are algebraic by $E_{\mathcal{A}}$. The set $E_{\mathcal{A}}$ is trivially closed under finite sums, and is easily seen to be closed under multiplication; consequently it is a ring.

Tarski [Tarski 1951] proved that the theory of the field of complex numbers was decidable; an effective (in the recursion-theoretic sense) procedure for converting formulas to a logically equivalent quantifier-free form was given. Consequences of this result include the existence of effective procedures for determining the number of distinct roots of a polynomial, and testing the equality of algebraic numbers defined by formulas.

We now demonstrate how to compute the probability of the set of paths which start at a designated state and satisfy a specified path formula. Consider a path formula of the form $\psi_0 U_{[a_1, b_1]} \psi_1 U_{[a_2, b_2]} \psi_2 \dots$

First, consider the case where the time intervals $[a_1, b_1], [a_2, b_2], \dots$ are non overlapping.

We define the following matrices.

—a transition rate matrix $Q_{i,i}$ obtained from Λ , that treats $\llbracket \psi_i \rrbracket_M^c$ as an absorbing set of states. This is obtained by using

$$\begin{aligned} q(j, k) &= \lambda_{j,k} \text{ if } s_j \in \llbracket \psi_i \rrbracket_M \\ &= 0 \text{ if } s_j \in \llbracket \psi_i \rrbracket_M^c \end{aligned}$$

this enables us to model the transitions where the Markov chain remains in $\llbracket \psi_i \rrbracket_M$.

—a transition rate matrix $Q_{i-1,i}$ obtained from Λ , that treats $\llbracket \psi_{i-1} \rrbracket_M^c \cap \llbracket \psi_i \rrbracket_M^c$ as an absorbing set of states. For this we use

$$\begin{aligned} q(j, k) &= \lambda_{j,k} \text{ if } s_j \in \llbracket \psi_i \rrbracket_M \cup \llbracket \psi_{i-1} \rrbracket_M \\ &= 0 \text{ if } s_j \in \llbracket \psi_i \rrbracket_M^c \cap \llbracket \psi_{i-1} \rrbracket_M^c \end{aligned}$$

this allows us to model the transitions from $\llbracket \psi_{i-1} \rrbracket_M$ to $\llbracket \psi_i \rrbracket_M$.

—An indicator matrix I_i for $\llbracket \psi_i \rrbracket_M$, such that

$$\begin{aligned} I_i(j, k) &= 1 \quad \text{if } s_j = s_k \in \llbracket \psi_i \rrbracket_M \\ &= 0 \quad \text{otherwise} \end{aligned}$$

Hence, the probability of a formula of the form

$$f_1 = \psi_0 U_{[a_1, b_1]} \psi_1 U_{[a_2, b_2]} \psi_2 \cdots U_{[a_n, b_n]} \psi_n \quad (1)$$

is given by

$$\begin{aligned} \mu^s(f_1) &= \pi_s \cdot P_{0,0}(a_1) \cdot I_0 \cdot P_{0,1}(b_1 - a_1) \cdot I_1 \cdot P_{1,1}(a_2 - b_1) \cdot \\ &I_1 \cdot P_{1,2}(b_2 - a_2) \cdot I_2 \cdots P_{n-1,n}(b_n - a_n) \cdot I_n \cdot \underline{1} \end{aligned} \quad (2)$$

where $P_{l,m}(t), t \geq 0$ is the one step transition matrix for time t corresponding to the transition rate matrix $Q_{l,m}$, π_s is the starting probability distribution, which in our case has unity for state s and zeros otherwise, and $\underline{1}$ is the column vector whose elements are all 1.

For a continuous time Markov chain with a transition rate matrix Q , the one step transition matrix for time t is given by $P(t) = e^{Qt}$. Note that entries of Q are rationals, and the arguments of $P_{i-1,i}$ are rationals (since a_i, b_i are rational). This observation leads to the following lemma:

LEMMA 1. *Each element of the $P_{l,m}(t)$ matrices as specified in Equation 2 may be expressed as a finite sum $\sum_j \eta_j e^{\delta_j t}$ where η_j and δ_j are algebraic complex numbers.*

PROOF. Let B be any square matrix whose entries are rationals. The matrix B can always be expressed in Jordan canonical form [Kaliath 1980], i.e., in the form $C \cdot J \cdot C^{-1}$. Here J is an upper block diagonal matrix as shown below:

$$\begin{bmatrix} J_1 & 0 & \cdots & 0 \\ 0 & J_2 & 0 & \cdots & 0 \\ 0 & \cdots & J_3 & \cdots & 0 \\ & & & \ddots & \\ 0 & \cdots & & & J_n \end{bmatrix}$$

The diagonal entries of each J_i are the eigenvalues of B , and the remaining entries of J_i are unity, as shown below:

$$\begin{bmatrix} \lambda_i & 1 & 0 & \cdots & 0 \\ 0 & \lambda_i & 1 & \cdots & 0 \\ & & & \ddots & \\ 0 & \cdots & 0 & & \lambda_i \end{bmatrix}$$

The size of J_i is equal to the multiplicity of λ_i . Since the eigenvalues are the solutions of the characteristic equation of B and the entries of B are rationals, the eigenvalues are, by definition, algebraic complex numbers. The entries of C and C^{-1} can be computed using linear algebra, and thus are also algebraic complex numbers.

The matrix e^{Bt} is equal to $C \cdot e^{Jt} \cdot C^{-1}$ and e^{Jt} is of the following form:

$$\begin{bmatrix} e^{J_1 t} & 0 & \dots & 0 \\ 0 & e^{J_2 t} & 0 & \dots & 0 \\ 0 & \dots & e^{J_3 t} & \dots & 0 \\ & & & \ddots & \\ 0 & \dots & & & e^{J_n t} \end{bmatrix}$$

The sub-matrix $e^{J_i t}$ is of the form

$$\begin{bmatrix} e^{\lambda_i t} & t e^{\lambda_i t} & (t^2 e^{\lambda_i t})/2! & \dots & (t^{m_i} e^{\lambda_i t})/(m_i)! \\ 0 & e^{\lambda_i t} & t e^{\lambda_i t} & \dots & (t^{m_i-1} e^{\lambda_i t})/(m_i-1)! \\ & & & \ddots & \\ & & & & e^{\lambda_i t} \end{bmatrix}$$

By inspection, the elements of $e^{J_i t}$ are members of $E_{\mathcal{A}}$. Since $E_{\mathcal{A}}$ is a ring, it is closed under products and finite sums. Hence the lemma follows. \square

Applying Lemma 1 to Equation 3 we see that $\mu^s(f_1)$ is a member of $E_{\mathcal{A}}$, i.e., equal to an expression of the form $\sum_k \eta_k e^{\delta_k}$ where the η_k, δ_k are algebraic. Since there are effective procedures for checking the equality of algebraic numbers, $\mu^s(f_1)$ can be effectively simplified to an expression of the form $\sum_{k'} \eta_{k'} e^{\delta_{k'}}$ where the $\eta_{k'}$'s are non zero, and the $\delta_{k'}$'s are distinct.

In general, it is extremely difficult to verify relationships between transcendental numbers [Richardson 1997]; skeptics are invited to check if $e^{\pi\sqrt{163}}$ is equal to 262537412640768744. Indeed, the theory of the real exponential field is conjectured to be decidable, but no complete proof exists of this fact [Wilkie 1995].

However, we can effectively decide if $\mu^s(f_1) > c$, by exploiting a celebrated theorem of transcendental number theory [Niven 1956].

THEOREM 2 (LINDEMANN-WEIERSTRASS). *Let c_1, \dots, c_n be pairwise distinct algebraic complex numbers. Then there exists no equation $a_1 e^{c_1} + \dots + a_n e^{c_n} = 0$ in which a_1, \dots, a_n are algebraic numbers and are not all zero.*¹

Suppose the expression $\sum_{k'} \eta_{k'} e^{\delta_{k'}}$ is degenerate, i.e., it consists of a single term of the form η_0 . Then the expression denotes an algebraic number, and it can be effectively checked if it is greater than c .

If it is not degenerate, invoking the Lindemann-Weierstrass theorem and noting that c is rational, we see that $\mu^s(f_1)$ can not be equal to c and so $|\mu^s(f_1) - c| > 0$.

Decidability of model checking follows from the following lemma.

LEMMA 2. *Given a transcendental real r of the form $\sum_j \eta_j e^{\delta_j}$ where the η_j and δ_j are algebraic complex numbers, and the δ_j 's are pairwise distinct, there is an effective procedure to test if $r > c$ for rational c .*

PROOF. Suppose a sequence of algebraic numbers S_1, S_2, \dots such that $|r - S_k| < 2^{-k}$ can be effectively constructed. Let $|r - c| = a > 0$. By the triangle inequality, $|r - c| \leq |r - \text{Re}(S_k)| + |\text{Re}(S_k) - c|$. Hence $|r - \text{Re}(S_k)| + |\text{Re}(S_k) - c|$ is

¹This result implies the transcendence of π (take $n = 2$, $c_1 = 0, c_2 = i\pi$); it was the first proof of the nonalgebraic nature of π . For a highly readable account of the development of this theorem, refer to [Ewing 1991].

bounded away from 0 by a . Since r is real, $|r - \operatorname{Re}(S_k)| \leq |r - S_k| < 2^{-k}$, and $|r - \operatorname{Re}(S_k)| + |\operatorname{Re}(S_k) - c|$ is bounded away from 0 by a , for sufficiently large k , it must be that $|\operatorname{Re}(S_k) - c| > 2^{-k}$. The sign of $\operatorname{Re}(S_k) - c$ is the sign of $r - c$.

In order to construct the sequence S_1, S_2, \dots we use the fact that e^z can be approximated with an error of less than ϵ (when $\epsilon < 1$) by taking the first $\lceil (3 \cdot |z|^2 / \epsilon) \rceil + 1$ terms of the Maclaurin expansion for e^z . This can be extended to obtain an upper bound on the number of terms needed to approximate r to within ϵ . Since the individual terms in the Maclaurin expansion are algebraic functions of the δ_j 's, it follows that the approximations are algebraic. \square

Now consider the case in which the successive intervals $[a_i, b_i]$ where the transitions are desired are allowed to overlap. Since a formula is finite, we can have a finite number of overlapping intervals. A key observation is that the finite number of overlaps allows us to partition the time in a finite number of nonoverlapping intervals and write the probability of the specification (set of acceptable paths) as a sum of the probabilities of disjoint events. This enables us to write $\mu^s(f_1)$ as the sum of exponentials of algebraic complex numbers, weighted by algebraic coefficients. To illustrate this, consider the formula

$$f_2 = \psi_0 U_{[a_1, b_1]} \psi_1 U_{[a_2, b_2]} \psi_2$$

where $0 < a_1 < a_2 < b_1 < b_2$. In this case, we may realize f_2 as one of four disjoint cases and hence we can write

$$\begin{aligned} \mu^s(f_2) = & \mu^s(\psi_0 U_{[a_1, a_2]} \psi_1 U_{[a_2, b_1]} \psi_2) + \mu^s(\psi_0 U_{[a_1, a_2]} \psi_1 U_{[b_1, b_2]} \psi_2) \\ & + \mu^s(\psi_0 U_{[a_2, b_1]} \psi_1 U_{[b_1, b_2]} \psi_2) + \mu^s(\psi_0 U_{[a_2, b_1]} \psi_1 U_{[a_2, b_1]} \psi_2) \end{aligned} \quad (3)$$

The first three terms are equivalent to the case with nonoverlapping intervals. The last term involves having both the $\llbracket \psi_0 \rrbracket_M \rightarrow \llbracket \psi_1 \rrbracket_M$ and $\llbracket \psi_1 \rrbracket_M \rightarrow \llbracket \psi_2 \rrbracket_M$ transitions in the same interval $[a_2, b_1]$ in the correct order. This may be evaluated by integrating the probabilities over the time of the first transition.

$$\mu^s(\psi_0 U_{[a_2, b_1]} \psi_1 U_{[a_2, b_1]} \psi_2) = \pi_s P_{0,0}(a_2) I_0 \int_{a_2}^{b_1} P_{0,0}(t - a_2) I_0 Q_{0,1} I_1 P_{1,2}(b_1 - t) I_2 dt$$

It is clear that since the integrand involved algebraic terms and exponentials in algebraic complex numbers and t , the definite integral with rational limits can be written in the form of a sum of exponentials of algebraic numbers with algebraic coefficients. Hence, this term is in $E_{\mathcal{A}}$. The other three terms in Equation 3 correspond to forms equivalent to the nonoverlapping intervals case, and hence already satisfy the decidability criteria.

5. CONCLUSIONS AND FUTURE WORK

We have defined a logic for specifying properties of finite state continuous time Markov chains. The model checking problem for this logic was shown to be decidable through a combination of results in algebraic and transcendental number theory.

In practice, we expect that rational approximations to the expression on the right hand side of Equation 3 as computed by standard numerical methods should suffice. Baier *et al.* [Baier et al. 1999] describe an implementation of such a model checker;

by clever choice of data structures, their procedure avoids some of the complexity of verifying systems which result from the composition of individual Markov chains.

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