

# Equivalence of Continuous-Time Markov Chains and Linear Dynamical Systems

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## Abstract

The purpose of this short note is to record that an analogue of the following result, which is known for discrete-time linear dynamical systems, also holds in the continuous-time setting. The dynamics of a  $d$ -state Markov chain is governed by that of a linear dynamical system of dimension at most  $d - 1$ ; conversely, a linear dynamical system of dimension  $d - 1$  can be “embedded” into a Markov chain with  $d$  states.

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## 1 Statements

A discrete-time linear dynamical system of dimension  $d$  is given by an update matrix  $M \in \mathbb{Q}^{d \times d}$  and an initial vector  $u_0 = u \in \mathbb{Q}^d$ . The system evolves as  $u_{n+1} = Mu_n$ , and hence  $u_n = M^n u_0$ . We say that this system is a (discrete) Markov chain if (i) the initial vector  $u_0$  is a distribution, i.e., all entries are non-negative, and the sum of entries is 1, and (ii) each column of  $M$  is a distribution, making it a stochastic matrix. We immediately observe that the orbit is thus a sequence of distributions.

Analogously, a continuous-time linear dynamical system of dimension  $d$  is given by an infinitesimal generator (or simply generator) matrix  $A \in \mathbb{Q}^{d \times d}$  and an initial vector  $v(0) \in \mathbb{Q}^d$ . The system evolves as  $\frac{d}{dt}v(t) = Av(t)$ , and hence  $v(t) = \exp(At)v(0)$ . We say that this system is a continuous Markov chain if (i) the initial vector  $v(0)$  is a distribution as before, and (ii) the entries in each column of the generator matrix  $A$  add up to 0, and the off-diagonal entries are non-negative. We refer to such a matrix  $A$  as a stochastic generator matrix.

We prove analogues of the core results in [1] (see also [2]). For notational convenience, we use  $\mathbf{0}_d$  to denote the  $d$ -dimensional vector with all entries 0,  $\mathbf{1}_d$  to denote the  $d$ -dimensional vector with all entries 1, and  $I_d$  to denote the  $d \times d$  identity matrix. We may omit the subscripts when the dimensions are clear from the context. We use  $O$  to denote a block of appropriate dimensions, all of whose entries are 0. The symbol  $\top$ , when in superscript, denotes matrix transposition. Linear algebraic notions that we assume familiarity with can be found in [1, Sec. 2.3].

► **Theorem 1.** *Given a distribution  $v \in \mathbb{Q}^d$  and a stochastic generator matrix  $A \in \mathbb{Q}^{d \times d}$ , let  $\ell \geq 1$  be the multiplicity of the eigenvalue 0 of  $A$ . We can compute a generator matrix  $B \in \mathbb{Q}^{(d-\ell) \times (d-\ell)}$ , an embedding matrix  $Q \in \mathbb{Q}^{d \times (d-\ell)}$ , a distribution  $s \in \mathbb{Q}^d$ , and a vector  $u \in \mathbb{Q}^{d-\ell}$  such that for all  $t \geq 0$ , we have that*

$$\exp(At)v = s + Q \exp(Bt)u.$$

*In the above, the choice of  $Q$  is independent of  $v$ .*

► **Theorem 2.** *Given a vector  $u \in \mathbb{Q}^{d-1}$  and a generator matrix  $C \in \mathbb{Q}^{(d-1) \times (d-1)}$ , we can compute  $\rho, \eta \in \mathbb{Q}$ , a distribution  $v \in \mathbb{Q}^d$ , and a stochastic generator matrix  $A \in \mathbb{Q}^{d \times d}$  such*



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that for all  $t \geq 0$  we have that

$$\exp(At)v = s + \eta e^{-\rho t} \cdot Q \exp(Ct)u,$$

where the pair  $s, Q$  is arbitrarily chosen as  $s = \mathbf{1}_d/d$  and  $Q = \begin{bmatrix} I_{d-1} \\ -\mathbf{1}_{d-1}^\top \end{bmatrix}$ .

### 2 Proofs

We begin by noting a basic property of generator matrices, which resolves a potential ambiguity in the statement of Thm. 1 regarding the multiplicity of the eigenvalue 0.

► **Lemma 3.** *Let  $A$  be a stochastic generator matrix. We have that  $A$  has 0 as an eigenvalue, and its algebraic multiplicity is equal to its geometric multiplicity. Furthermore, all other eigenvalues of  $A$  have negative real part.*

**Proof.** It follows by definition that  $\mathbf{1}^\top A = \mathbf{0}^\top$ , and we thus have found a left eigenvector for the eigenvalue 0. Observe that we can choose  $a > 0$  small enough so that  $M = I + aA$  is a stochastic matrix. It is easy to check that  $\lambda$  is an eigenvalue of  $A$  if and only if  $\lambda' = 1 + a\lambda$  is an eigenvalue of  $M$ , and that their corresponding eigenvectors have the same order, and span the same spaces. Perron-Frobenius theory (see, e.g., [1, Lem. 3]) gives us the desired properties of the spectrum of  $M$ : it necessarily has 1 as an eigenvalue with equal algebraic and geometric multiplicities, and all other eigenvalues have absolute value at most 1. The former corresponds to the eigenvalue 0 of  $A$ , the latter can only correspond to eigenvalues of  $A$  that are strictly to the left of the imaginary axis. ◀

The proofs of Thms. 1 and 2 mirror the proofs in [1, Sec. 3].

**Proof of Thm. 1.** Let  $V_0$  be the  $\ell$ -dimensional subspace spanned by the eigenvectors of the eigenvalue 0, and let  $W$  be the  $(d - \ell)$ -dimensional subspace (of  $\mathbb{C}^d$ ) spanned by the (generalised) eigenvectors of all the other eigenvalues. Recall the basic properties that  $V_0, W$  together span  $\mathbb{C}^d$  [1, Thm. 6], and that  $W$  is perpendicular to the space spanned by the left eigenvectors of the eigenvalue 0.

We can thus compute matrices  $P \in \mathbb{Q}^{d \times \ell}, Q \in \mathbb{Q}^{d \times (d-\ell)}$  whose columns respectively form bases of  $V_0, W$ . Observe that  $v$  can be uniquely expressed as  $Pw + Qu$ , where  $w \in \mathbb{Q}^\ell$ , and  $u \in \mathbb{Q}^{d-\ell}$ . We choose  $s = Pw$ .

It remains to define  $B$ , and then show that  $\exp(At)v = s + Q \exp(Bt)u$ . To that end, define the invertible  $R = \begin{bmatrix} P & Q \end{bmatrix}$ , and express  $R^{-1}$  as  $\begin{bmatrix} P' \\ Q' \end{bmatrix}$ . Consider the matrix

$$\begin{aligned} D &= R^{-1}AR \\ &= \begin{bmatrix} P' \\ Q' \end{bmatrix} A \begin{bmatrix} P & Q \end{bmatrix} \\ &= \begin{bmatrix} P' \\ Q' \end{bmatrix} \begin{bmatrix} O & AQ \end{bmatrix} \\ &= \begin{bmatrix} O & O \\ O & Q'AQ. \end{bmatrix} \end{aligned}$$

In the above, we used the property that all of the columns of  $AQ$  are in  $W$ , and that the rows of  $P'$  span the space perpendicular to  $W$ . We define  $B = Q'AQ$ , and observe that by

similar reasoning as above, for all  $n \geq 1$

$$D^n = \begin{bmatrix} O & O \\ O & B^n \end{bmatrix} = \begin{bmatrix} O & O \\ O & Q'A^nQ \end{bmatrix}.$$

Dually, we also deduce that for all  $n \geq 1$ , we have  $A^n = QB^nQ'$ .

Let us now express the matrix exponential using the formal power series. We get

$$\begin{aligned} \exp(At)v &= R(\exp(Dt))R^{-1}v \\ &= [P \quad Q] \left( \sum_{i=0}^{\infty} \frac{t^i}{i!} D^i \right) \begin{bmatrix} w \\ u \end{bmatrix} \\ &= s + Qu + \sum_{i=1}^{\infty} \frac{t^i}{i!} [P \quad Q] \begin{bmatrix} O & O \\ O & B^i \end{bmatrix} \begin{bmatrix} w \\ u \end{bmatrix} \\ &= s + Qu + \sum_{i=1}^{\infty} \frac{t^i}{i!} QB^i u \\ &= s + Q \left( \sum_{i=0}^{\infty} \frac{t^i}{i!} B^i \right) u \\ &= s + Q \exp(Bt)u, \end{aligned}$$

as desired. ◀

We remark that [1, Sec. 3] also discusses how the spectral decompositions of  $B$  and  $A$  are related. The same discussion applies in the present setting too. In particular,  $B$  and  $A$  have the same nonzero eigenvalues, and hence  $Q \exp(Bt)u$  converges to  $\mathbf{0}$ .

**Proof of Thm. 2.** We prove the theorem by using the calculations from the previous proof. We have fixed  $Q$ , and  $P$  is the matrix whose single column is  $s$ . This determines  $P', Q'$ ; in particular  $P' = \mathbf{1}^\top$ . We choose the scalar  $\eta$  to ensure that  $v = s + \eta Qu$  is a distribution. The scalar  $\rho$  is chosen to define a matrix  $B = C - \rho I$  such that  $QBQ'$  is a stochastic generator matrix. It would then follow that

$$\exp(At)v = s + \eta Q \exp(Bt)u = s + \eta e^{-\rho t} \exp(Ct)u.$$

Let us show how to choose  $\rho$ . We already have that  $\mathbf{1}^\top Q C Q' = \mathbf{0}^\top$  because  $\mathbf{1}^\top Q = \mathbf{0}^\top$ . We only need to show that we can make the off-diagonal entries non-negative by subtracting  $\rho I$  from  $C$ .

We have

$$Q(C - \rho I)Q' = QCQ' - \rho QQ'.$$

Since

$$[s \quad Q] \begin{bmatrix} \mathbf{1}^\top \\ Q' \end{bmatrix} = I,$$

we have that  $QQ' = I - S$ , where  $S$  is the matrix with all entries equal to  $1/d$ . Thus,  $Q(C - \rho I)Q' = QCQ' + \rho S - \rho I$ . Choosing  $\rho$  to be large enough thus ensures that all the off-diagonal entries of  $QBQ'$  are non-negative, as desired. ◀

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**References**

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