

MATH180C: Introduction to Stochastic Processes II

Lecture A00: math-old.ucsd.edu/~ynemish/teaching/180cA

Lecture B00: math-old.ucsd.edu/~ynemish/teaching/180cB

Today: General continuous time
Markov chains. Matrix exponentials

Next: PK 6.3, 6.6, Durrett 4.2

Week 3:

- homework 2 (due Friday April 15)

Q-matrices (infinitesimal generators)

Let $S = \{0, 1, \dots, N\}$. We call $Q = (q_{ij})_{i,j=0}^N$ a Q-matrix if Q satisfies the following conditions:

(a) $0 \leq -q_{ii} < \infty$ for all i

$$q_i := \sum_{j \neq i} q_{ij}$$

(b) $q_{ij} \geq 0$ for all $i \neq j$

$$\text{then } q_{ii} = -q_i$$

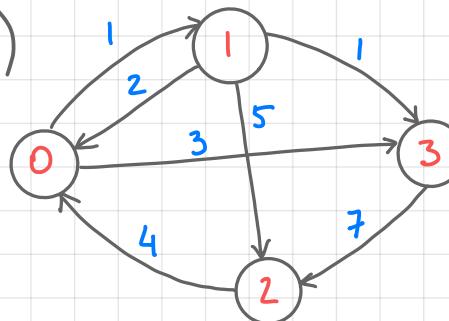
(c) $\sum_j q_{ij} = 0$ for all i

Examples

(a)

$$Q = \begin{pmatrix} -2 & 1 & 1 \\ 2 & -7 & 5 \\ 0 & 2 & -2 \end{pmatrix}$$

(b)



$$Q = \begin{pmatrix} 0 & 1 & 2 & 3 \\ 0 & -4 & 1 & 0 & 3 \\ 1 & 2 & -8 & 5 & 1 \\ 2 & 4 & 0 & -4 & 0 \\ 3 & 0 & 0 & 7 & -7 \end{pmatrix}$$

Matrix exponentials

Let $Q = (q_{ij})_{i,j=1}^N$ be a matrix. Then the series
 $\sum_{k=0}^{\infty} \frac{Q^k}{k!}$ converges componentwise, and we denote

its sum $\sum_{k=0}^{\infty} \frac{Q^k}{k!} =: e^Q$, the matrix exponential of Q .

In particular, we can define $e^{tQ} = \sum_{k=0}^{\infty} \frac{Q^k t^k}{k!}$ for $t \geq 0$.

Thm. Define $P(t) = e^{tQ}$. Then

$$(i) \quad P(t+s) = P(t)P(s) \quad \text{for all } s, t > 0$$

(ii) $(P(t))_{t \geq 0}$ is the unique solution to the equations

$$\begin{cases} \frac{d}{dt} P(t) = P(t)Q \\ P(0) = I \end{cases}, \quad \text{and} \quad \begin{cases} \frac{d}{dt} P(t) = QP(t) \\ P(0) = I \end{cases}$$

$$\sum_{k=0}^{\infty} \frac{t^k Q^k}{k!} Q = Q \sum_{k=0}^{\infty} \frac{t^k Q^k}{k!}$$

Matrix exponentials

Properties are easy to remember → scalar exponential

$$(i) e^{(t+s)Q} = e^{tQ} e^{sQ} = e^{sQ} e^{tQ} \quad (e^{(t+s)\alpha} = e^{t\alpha} e^{s\alpha})$$

(note that in general $AB \neq BA$ for matrices A, B)

$$(ii) \frac{d}{dt} e^{tQ} = Q e^{tQ} = e^{tQ} Q \quad \left(\frac{d}{dt} e^{t\alpha} = \alpha e^{t\alpha} \right)$$

$$e^{0 \cdot Q} = I \quad (e^0 = 1)$$

Example

$$(a) Q_1 = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \quad Q_1^2 = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \Rightarrow e^{tQ_1} = I + Q_1 t + \frac{Q_1^2}{2!} t^2 + \dots = \begin{pmatrix} 1 & t \\ 0 & 1 \end{pmatrix}$$

$$(b) Q_2 = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}, \quad e^{tQ_2} = \begin{pmatrix} e^{\lambda_1 t} & 0 \\ 0 & e^{\lambda_2 t} \end{pmatrix}$$

Matrix exponentials

Results on the previous slide hold for any matrix Q .

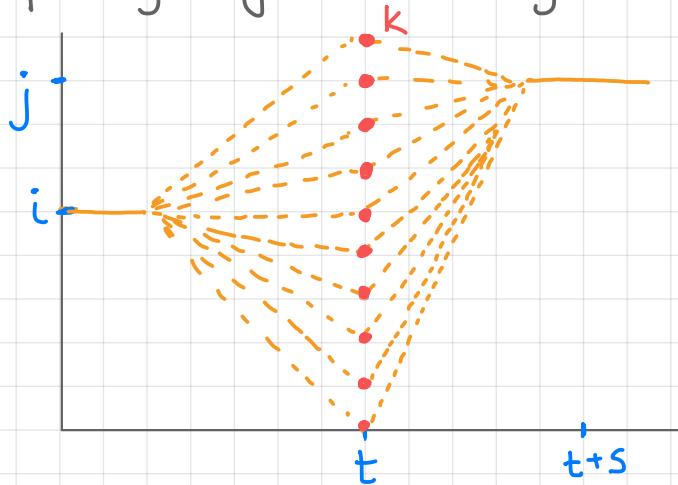
Thm. Matrix Q is a Q -matrix

iff $P(t) = e^{tQ}$ is a stochastic matrix $\forall t$
 $\sum_j P_{ij}(t) = 1$ for all i and $t \geq 0$

Remarks The semigroup property gives entrywise

$$P_{ij}(t+s) = [P(t)P(s)]_{ij}$$
$$= \sum_{k=0}^N P_{ik}(t) P_{kj}(s)$$

(if you think about MC \rightarrow
Chapman-Kolmogorov)



Main theorem

Let $P(t)$ be a matrix-valued function $t \geq 0$.

Consider the following properties

(a) $P_{ij}(t) \geq 0$, $\sum_j P_{ij}(t) = 1$ for all $i, j, t \geq 0$

(b) $P(0) = I$

(c) $P(t+s) = P(t)P(s)$ for all $t, s \geq 0$

(d) $\lim_{t \downarrow 0} P(t) = I$ (continuous at 0)

Theorem A. $P(t)$ satisfies (a)-(d)

if and only if

$$P(t) = e^{tQ} \text{ for some Q-matrix } Q$$

Main theorem. Remarks

This theorem establishes one-to-one correspondance between matrices $P(t)$ satisfying (a)-(d) and the Q -matrices of the same dimension.

Remarks

1. Conditions (a)-(d) imply that $P(t)$ is differentiable

2. If $P(t) = e^{tQ}$, then $P(h) = I + Qh + o(h)$ as $h \rightarrow 0$

$$P(h) = I + Qh + \sum_{k \geq 2} \frac{Q^k h^k}{k!} = o(h)$$

Q-matrices and Markov chains

Let $(X_t)_{t \geq 0}$ be a continuous time MC, $X_t \in \{0, 1, \dots, N\}$
with right-continuous sample paths

Denote $P_{ij}(t) = P(X_t = j | X_0 = i)$, $i, j \in \{0, 1, \dots, N\}$
stationary

Then

- $P_{ij}(t) \geq 0$, $\sum_{j=0}^N P_{ij}(t) = 1$ ($= \sum_{j=0}^N P(X_t = j | X_0 = i)$)
- $P_{ij}(0) = \delta_{ij}$ ($P(X_0 = j | X_0 = i) = \begin{cases} 1, & j = i \\ 0, & j \neq i \end{cases}$) $\Leftrightarrow P(0) = I$
- $P_{ij}(t+s) = P(X_{t+s} = j | X_0 = i) = \sum_{k=0}^N P_{kj}(s) P_{ik}(t)$
 $= \sum_{k=0}^N P(X_{t+s} = j | X_0 = i, X_t = k) P(X_t = k | X_0 = i)$
- $\lim_{h \rightarrow 0} P(X_h = j | X_0 = i) = \delta_{ij}$  $P(h) \rightarrow I, h \rightarrow 0$

Q-matrices and Markov chains (cont.)

$P(t)$ satisfies properties (a)-(d) from Theorem A.

\Rightarrow there is a Q-matrix Q such that

$$P(t) = e^{tQ}$$

$$P_{ij}(h) = q_{ij}h + o(h) \quad i \neq j$$

In particular,

$$P_{ii}(h) = 1 + q_{ii}h + o(h)$$

$$P(h) = I + Qh + o(h) \quad \text{as } h \rightarrow 0$$

This implies the one-to-one correspondance between Q-matrices and continuous time MC with right-continuous sample paths.

Q is called the infinitesimal generator of $(X_t)_{t \geq 0}$

Infinitesimal description of cont. time MC

Let $Q = (q_{ij})_{i,j=0}^N$ be a Q -matrix, let $(X_t)_{t \geq 0}$ be right-continuous stochastic process, $X_t \in \{0, 1, \dots, N\}$.

We call $(X_t)_{t \geq 0}$ a Markov chain with generator Q , if

(i) $(X_t)_{t \geq 0}$ satisfies the Markov property

$$(ii) P(X_{t+h} = j | X_t = i) = \begin{cases} q_{ij}h + o(h), & i \neq j \\ 1 + q_{ii}h + o(h), & i = j \end{cases} \text{ as } h \rightarrow 0$$

Example

Pure death process

- $P_{i,i-1}(h) = \mu_i h + o(h)$
- $P_{ii}(h) = 1 - \mu_i h + o(h)$
- $P_{ij}(h) = o(h) \text{ for } j \notin \{i-1, i\}$

The corresponding Q -matrix

$$Q = \begin{pmatrix} 0 & 0 & 0 & \cdots & \cdots \\ \vdots & \mu_1 & -\mu_1 & 0 & \cdots & \cdots \\ 2 & 0 & \mu_2 & -\mu_2 & 0 & \cdots & \cdots \\ 3 & & & \ddots & \ddots & \ddots & \ddots \\ \vdots & & & & \ddots & \ddots & \ddots \\ N & & & & & \ddots & \mu_N - \mu_N \end{pmatrix}$$

Sojourn time description

Let $Q = (q_{ij})_{i,j=0}^N$ be a Q-matrix. Denote $q_i = \sum_{j \neq i} q_{ij}$

so that

$$Q = \begin{pmatrix} -q_0 & q_{01} & q_{02} & \cdots \\ q_{10} & -q_1 & q_{12} & \cdots \\ q_{20} & q_{21} & -q_2 & \cdots \\ \vdots & \vdots & \ddots & \vdots \end{pmatrix} \quad q_0 = \sum_{i \neq 0} q_{0i}$$

Denote $Y_k := X_{W_k}$ (jump chain).

Then the MC with generator matrix Q has the following equivalent jump and hold description

- sojourn times S_k are independent r.v.

with $P(S_k > t \mid Y_k = i) = e^{-q_it} \quad (S_k \sim \text{Exp}(q_i))$

- transition probabilities $P(Y_{k+1} = j \mid Y_k = i) = \frac{q_{ij}}{q_i}$