

MATH180C: Introduction to Stochastic Processes II

<https://mathweb.ucsd.edu/~ynemish/teaching/180c>

Today: General CTMC. Matrix
FSA for general MC

Next: PK 6.3, 6.6, Durrett 4.2

Week 3:

- HW2 due Friday, April 21 on Gradescope
- No in-person lecture on Friday, April 21

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 $\forall t$

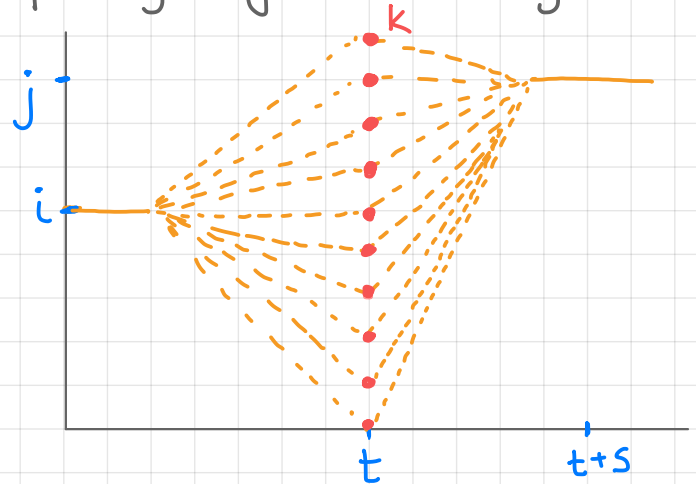
$$P_{ij}(t) \geq 0, \quad \sum_j P_{ij}(t) = 1 \quad \forall i \text{ and } \forall t \geq 0$$

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) $P(0) = I = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix}$

(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

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

Main theorem. Remarks

This theorem establishes a 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 $h \rightarrow 0$


$$P(h) = I + hQ + \sum_{k=2}^{\infty} \frac{Q^k h^k}{k!} \quad o(h)$$

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

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

Q-matrices and Markov chains (cont.)

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

\Rightarrow there is a Q-matrix Q such that

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

In particular,

$$P(h) = I + Qh + o(h)$$

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

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

Example

Pure death process

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

The corresponding Q -matrix

$$Q = \begin{matrix} 0 \\ 1 \\ 2 \\ \vdots \\ N \end{matrix} \begin{pmatrix} 0 & 0 & 0 & \dots & 0 \\ \mu_1 & -\mu_1 & 0 & \dots & 0 \\ 0 & \mu_2 & -\mu_2 & \dots & 0 \\ & & & \ddots & \\ 0 & \dots & \dots & & 0 \mu_N - \mu_N \end{pmatrix}$$

Sojourn time description

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

so that

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

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

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

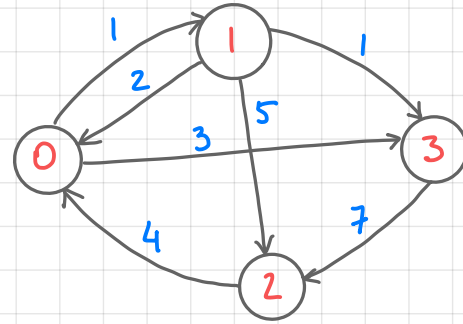
- sojourn times S_k are independent r.v.

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

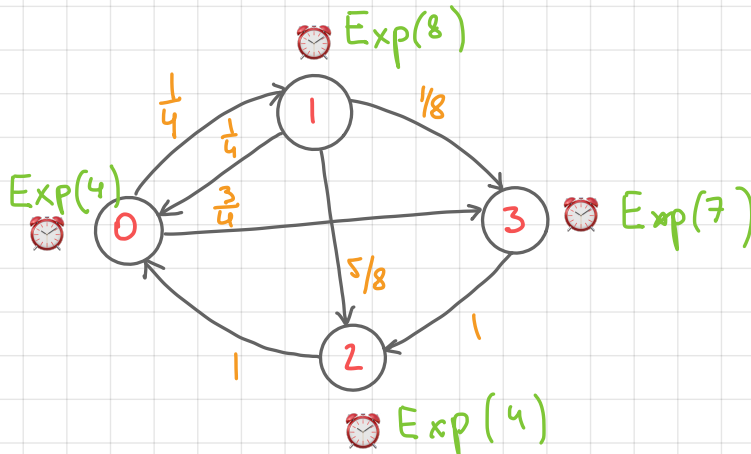
- transition probabilities $P(Y_{k+1} = j \mid Y_k = i) = \frac{q_{ij}}{q_i}$

Example

	0	1	2	3
0	-4	1	0	3
1	2	-8	5	1
2	4	0	-4	0
3	0	0	7	-7

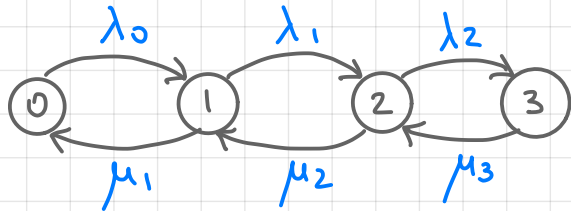


$i \xrightarrow{\alpha} j \text{ " = " } P_{ij}(h) = \alpha h r_0(h)$



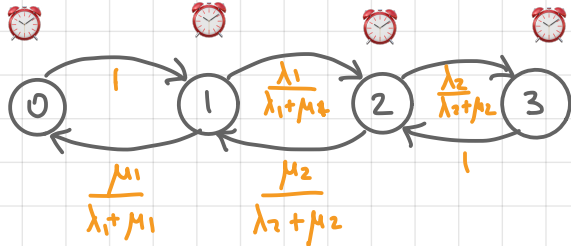
Example

Birth and death process on $\{0, 1, 2, 3\}$



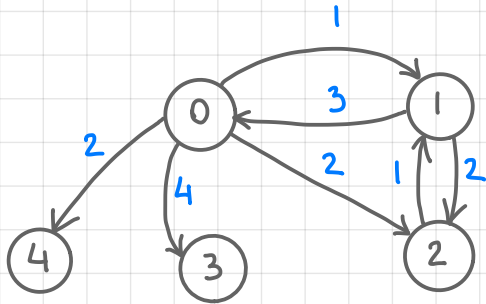
$$Q = \begin{pmatrix} -\lambda_0 & \lambda_0 & & \\ \mu_1 & -(\lambda_1 + \mu_1) & \lambda_1 & \\ & \mu_2 & -(\lambda_2 + \mu_2) & \lambda_2 \\ & & \mu_3 & -\mu_3 \end{pmatrix}$$

$\text{Exp}(\lambda_0)$ $\text{Exp}(\lambda_1 + \mu_1)$ $\text{Exp}(\lambda_2 + \mu_2)$ $\text{Exp}(\mu_3)$



General continuous time finite state MCs

Rate diagram



Generator

$$Q = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{pmatrix} -9 & 1 & 2 & 4 & 2 \\ 3 & -5 & 2 & & \\ & 1 & -1 & & \\ & & & 0 & \\ & & & & 0 \end{pmatrix} \end{matrix}$$

Infinitesimal description

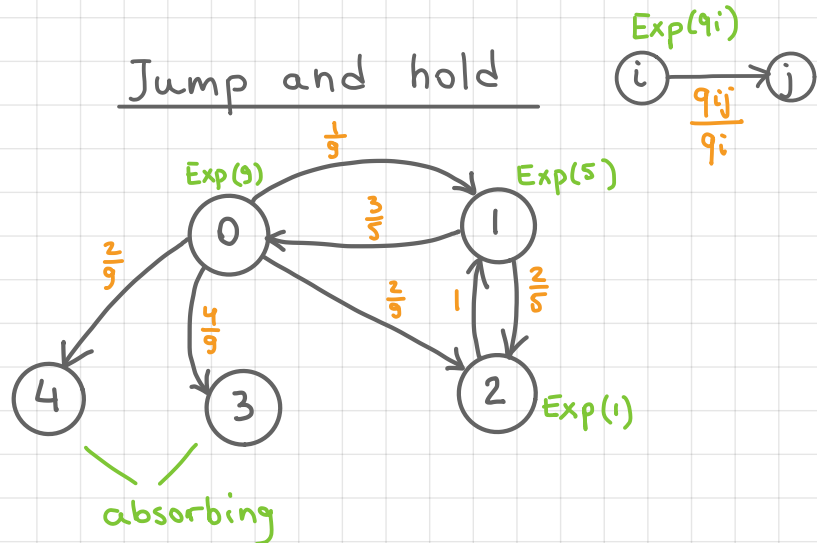
$$P_{ij}(h) = q_{ij}h + o(h), \quad i \neq j$$

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

$$P_{02}(h) = 2h + o(h)$$

$$P_{00}(h) = 1 - 9h + o(h)$$

Jump and hold



Absorption probabilities for finite state chains

By considering the jump chain $(Y_n)_{n \geq 0}$ with $Y_n = X_{W_n}$ and its transition probabilities $P(Y_{n+1}=j | Y_n=i) = \frac{q_{ij}}{q_i}$ we can apply the first step analysis to compute, e.g., the absorption probabilities (similarly as for B&D)

If state i is absorbing, then $q_{ij} = 0$ for all $j \neq i$ (no jumps from state i), so $q_i = q_{ii} = 0$. Let Q be given by

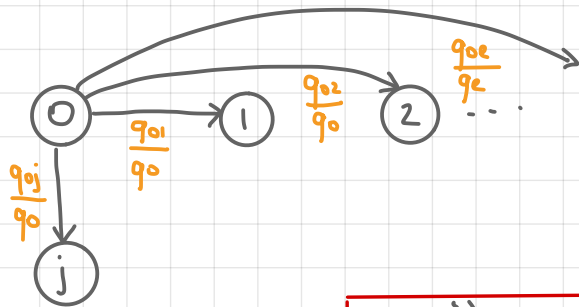
$$Q = \begin{array}{c} 0 \\ \vdots \\ k-1 \\ k \\ \vdots \\ N \end{array} \left(\begin{array}{ccc|ccc} 0 & \dots & k-1 & k & \dots & N \\ -q_0 & & & q_{ij} & & \\ \vdots & & & \vdots & & \\ q_{ij} & \dots & -q_{k-1} & & & \\ \hline & & 0 & & & 0 \\ & & & & \ddots & \\ & & & & & 0 \end{array} \right)$$

with $\{0, \dots, k-1\}$ transient,
 $\{k, \dots, N\}$ absorbing

Absorption probabilities for finite state chains

$$Q = \begin{matrix} 0 \\ \vdots \\ k-1 \\ k \\ \vdots \\ N \end{matrix} \left(\begin{array}{ccc|ccc} 0 & \dots & k-1 & k & \dots & N \\ \hline -q_0 & & & q_{ij} & & \\ \vdots & & & \vdots & & \\ q_{ij} & \dots & -q_{k-1} & & & \\ \hline & & & 0 & & 0 \\ & & & \vdots & & \\ & & & & \dots & \\ & & & & & 0 \end{array} \right)$$

Jump chain



Let $i \in \{0, \dots, k-1\}$, $j \in \{k, \dots, N\}$.

Let $M = \min\{n: Y_n \in \{k, \dots, N\}\}$

Denote $u_i^{(j)} = P(Y_M = j | X_0 = i)$.

Then FSA leads to the system

$$u_i^{(j)} = P(Y_M = j | Y_0 = i)$$

=

$$u_i^{(j)} = \frac{q_{ij}}{q_i} + \sum_{\substack{c=0 \\ c \neq i}}^{k-1} \frac{q_{ic}}{q_i} u_c^{(j)}$$

$P(Y_{n+1} = j | Y_n = i)$

$P(Y_{n+1} = c | Y_n = i)$