

# MATH180C: Introduction to Stochastic Processes II

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

Today: Absorption times.

General CTMC. Matrix exponentials

Next: 6.6, Durrett 4.1

Week 3:

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

# Mean time until absorption

Let  $(X_t)_{t \geq 0}$  be a birth and death process. Denote

$T = \min\{t \geq 0 : X_t = 0\}$  absorption time and

Let  $(Y_n)_{n \geq 0}$  be the

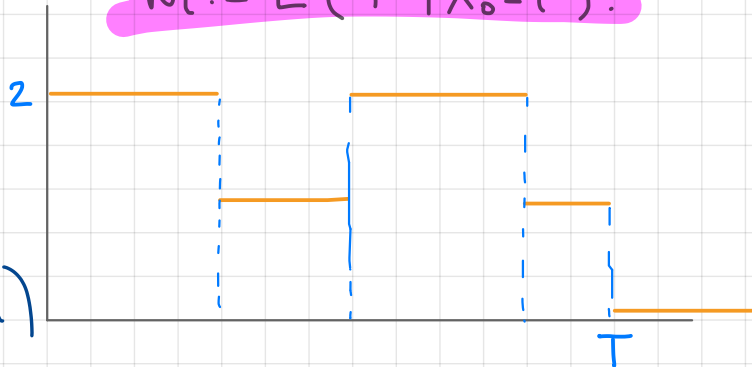
jump chain for  $(X_t)_{t \geq 0}$ .

$$N := \min\{n \geq 0 : Y_n = 0\}$$

$$\text{Then } T = \sum_{k=0}^{N-1} S_k$$

$$= E(S_0 | X_0 = i)$$

$$w_i := E(T | X_0 = i)$$



$$w_i = E\left(\sum_{k=0}^{N-1} S_k | X_0 = i\right) = \frac{1}{\lambda_i + \mu_i} + E\left(\sum_{k=1}^{N-1} S_k | X_0 = i\right)$$

$$= \frac{1}{\lambda_i + \mu_i} + \underbrace{E\left(\sum_{k=1}^{N-1} S_k | X_0 = i, Y_1 = i+1\right)}_{\text{SMP II } w_{i+1}} P(Y_1 = i+1 | Y_0 = i) + \underbrace{E\left(\sum_{k=1}^{N-1} S_k | X_0 = i, Y_1 = i-1\right)}_{\text{SMP } w_{i-1}} P(Y_1 = i-1 | Y_0 = i)$$

$\frac{\lambda_i}{\lambda_i + \mu_i}$  (under  $P(Y_1 = i+1 | Y_0 = i)$ ) and  $\frac{\mu_i}{\lambda_i + \mu_i}$  (under  $P(Y_1 = i-1 | Y_0 = i)$ )

## Mean time until absorption

$$\begin{cases} w_i = \frac{1}{\lambda_i + \mu_i} + \frac{\lambda_i}{\lambda_i + \mu_i} w_{i+1} + \frac{\mu_i}{\lambda_i + \mu_i} w_{i-1}, \\ w_0 = 0 \end{cases}$$

$$w_i = \begin{cases} \sum_{j=1}^{\infty} \frac{1}{\lambda_j p_j} + \sum_{k=1}^{i-1} p_k \sum_{j=k+1}^{\infty} \frac{1}{\lambda_j p_j}, & \text{if } \sum_{j=1}^{\infty} \frac{1}{\lambda_j p_j} < \infty \\ \infty, & \text{if } \sum_{j=1}^{\infty} \frac{1}{\lambda_j p_j} = \infty \end{cases}$$

# First step analysis for birth and death processes

Summary:

Let  $(X_t)_{t \geq 0}$  be a birth and death process of rates  $((\lambda_i, \mu_i))_{i \geq 0}$  with  $\lambda_0 = 0$  (state 0 absorbing).

Denote  $T = \min\{t : X_t = 0\}$ ,  $u_i = P(X_t \text{ gets absorbed in } 0 | X_0 = i)$

$w_i = E(T | X_0 = i)$  and  $p_j = \frac{\mu_1 \mu_2 \dots \mu_j}{\lambda_1 \lambda_2 \dots \lambda_j}$ . Then

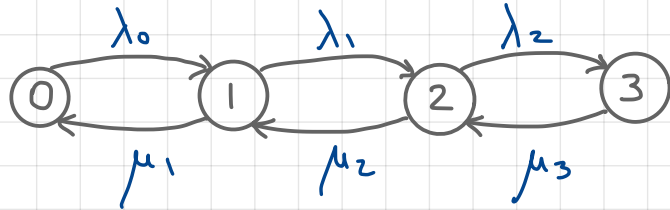
$$u_i = \begin{cases} \frac{\sum_{j=i}^{\infty} p_j}{1 + \sum_{j=i}^{\infty} p_j}, & \text{if } \sum_{j=1}^{\infty} p_j < \infty \\ 1, & \text{if } \sum_{j=1}^{\infty} p_j = \infty \end{cases}$$
$$w_i = \begin{cases} \sum_{j=1}^{\infty} \frac{1}{\lambda_j p_j} + \sum_{k=1}^{i-1} p_k \sum_{j=k+1}^{\infty} \frac{1}{\lambda_j p_j}, & \text{if } \sum_{j=1}^{\infty} \frac{1}{\lambda_j p_j} < \infty \\ \infty, & \text{if } \sum_{j=1}^{\infty} \frac{1}{\lambda_j p_j} = \infty \end{cases}$$

## Birth and death processes. Results

- infinitesimal transition probability description
- sojourn time description (jump and hold)  
sojourn times are independent exponential r.v.s  
$$P(i \rightarrow i+1) = \frac{\lambda_i}{\lambda_i + \mu_i}, \quad P(i \rightarrow i-1) = \frac{\mu_i}{\lambda_i + \mu_i}$$
- system of differential equations for pure birth/death  
e.g.  $P_i'(t) = -\lambda_i P_i(t) + \lambda_{i-1} P_{i-1}(t)$
- distributions of  $X_t$  for linear birth (geometric) and linear death (binomial) processes
- first step analysis giving absorption probabilities and mean time to absorption
- explosion, Strong Markov property etc.

# General continuous time MC

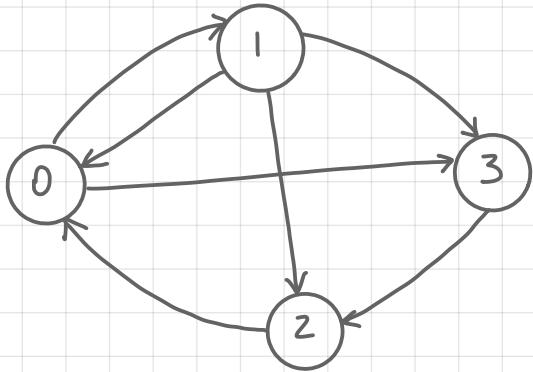
Assume for simplicity that the state space is finite  
 $(X_t)_{t \geq 0}$  is right-continuous



birth and death process

$$P_{i,i+1}(h) = \lambda_i h + o(h)$$

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



general MC

$$P_{ij}(s) := P(X_{t+s} = j \mid X_t = i) = P(X_s = j \mid X_0 = i)$$

$$i \neq j \quad P_{ij}(h) = q_{ij} \cdot h + o(h) \quad \text{as } h \downarrow 0$$

How to define? How to analyze?

# 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:

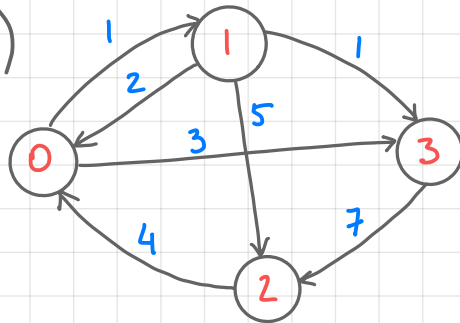
- (a)  $0 \leq -q_{ii} < \infty$  for all  $i$        $q_i := \sum_{j \neq i} q_{ij}$   
(b)  $q_{ij} \geq 0$  for all  $i \neq j$        $q_{ii} = -q_i$   
(c)  $\sum_j q_{ij} = 0$  for all  $i$

## Examples

(a)

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

(b)



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

# Matrix exponentials

$$Q^0 = I$$

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

**semigroup**

(i)  $P(t+s) = P(t)P(s)$  for all  $s, t > 0$

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

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



# Matrix exponentials

Properties are easy to remember  $\rightarrow$  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

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

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

# 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  $\forall t$

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

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)

