

Performance Evaluation of Networks

Sara Alouf

Ch 1 - Discrete-Time Markov Chain (DTMC)

- Properties of DTMC:

- ▶ aperiodicity, irreducibility, positive recurrence, ergodicity

- n -step transition matrix \mathbf{P}^n

- Transient distribution

$$\pi(n) = \pi(0) \mathbf{P}^n$$

- Limiting distribution

$$\lim_{n \rightarrow \infty} \pi(n) = \pi(0) \lim_{n \rightarrow \infty} \mathbf{P}^n$$

- Stationary distribution

$$\pi = \pi \mathbf{P}$$

$$\pi \mathbf{1} = 1$$

Ch 2 - Continuous-Time Markov Chain (CTMC)

- Stochastic process $\{X(t), t \geq 0\}$ with discrete state-space and verifying Markov property

$$P(X(t) = j \mid X(s_1) = i_1, \dots, X(s_{n-1}) = i_{n-1}, X(s) = i) \\ = P(X(t) = j \mid X(s) = i)$$

$$i_1, \dots, i_{n-1}, i, j \in \mathcal{E}$$

$$0 \leq s_1 < \dots < s_{n-1} < s < t$$

is a **continuous-time Markov chain**

- CTMC is homogeneous if

$$P(X(t) = j \mid X(s) = i) = p_{i,j}(t - s)$$

$$\forall i, j \in \mathcal{E}, 0 \leq s < t$$

Chapman-Kolmogorov Equation

- Proposition 4: For all $t > 0, s > 0, i, j \in \mathcal{E}$

$$p_{i,j}(t+s) = \sum_{k \in \mathcal{E}} p_{i,k}(t) p_{k,j}(s)$$

$$\mathbf{P}(t+s) = \mathbf{P}(t) \cdot \mathbf{P}(s)$$

DTMC:

$$p_{i,j}^{(n+m)} = \sum_{k \in \mathcal{E}} p_{i,k}^{(n)} p_{k,j}^{(m)}$$

$$\mathbf{P}^{(n+m)} = \mathbf{P}^{(n)} \mathbf{P}^{(m)}$$

$$\mathbf{P}^{(n)} = \mathbf{P}^n$$

- Proof: use law of total probability and Markov property
- Matrix notation

$$\mathbf{P}(t) = \left[p_{i,j}(t) \right]_{i,j \in \mathcal{E}}$$

- CTMC: real steps vs. DTMC: integer steps

Infinitesimal Generator

- Define

$$\left. \begin{aligned} q_{i,i} &:= \lim_{h \rightarrow 0} \frac{p_{i,i}(h) - 1}{h} \leq 0 \\ q_{i,j} &:= \lim_{h \rightarrow 0} \frac{p_{i,j}(h)}{h} \geq 0 \end{aligned} \right\} \quad q_{i,i} = - \sum_{j \neq i} q_{i,j}$$

- Matrix

$$\mathbf{Q} = [q_{i,j}] = \lim_{h \rightarrow 0} \frac{\mathbf{P}(h) - \mathbf{I}}{h} \quad \mathbf{I} = \begin{bmatrix} 1 & 0 & \cdots \\ 0 & 1 & \cdots \\ \vdots & \vdots & \ddots \end{bmatrix}$$

- If CTMC can take any integer value

$$\mathbf{Q} = \begin{pmatrix} -\sum_{j \neq 0} q_{0,j} & q_{0,1} & q_{0,2} & \cdots \\ q_{1,0} & -\sum_{j \neq 1} q_{1,j} & q_{1,2} & \cdots \\ \vdots & \vdots & \vdots & \vdots \end{pmatrix} \quad \sum = 0$$

Interpretation

- $q_{i,j}$ is the transition rate to state j when in state i
- $-q_{i,i}$ is transition rate out of state i
- Sojourn time in state i is $\text{Exp}(-q_{i,i})$ $P(S(i) > x) = e^{q_{i,i}x}$

Proof: assume $X(0) = i$ and sojourn time is $S(i)$

$$\begin{aligned} P(S(i) > x + h) &= P(S(i) > x \text{ and } X(t) = i, x < t \leq x + h) \\ &= P(S(i) > x) P(X(t) = i, x < t \leq x + h) \end{aligned}$$

When $h \rightarrow 0$, $P(X(t) = i, x < t \leq x + h) \approx p_{i,i}(h)$

$$q_{i,i} = \lim_{h \rightarrow 0} \frac{p_{i,i}(h) - 1}{h} \Rightarrow p_{i,i}(h) = 1 + hq_{i,i} + o(h)$$

Proof of Exp Sojourn Time (continued)

$$P(S(i) > x + h) - P(S(i) > x) = P(S_i > x)(hq_{i,i} + o(h))$$

Divide by h and take limit as $h \rightarrow 0$

$$\frac{dP(S(i) > x)}{dx} = q_{i,i} P(S(i) > x)$$

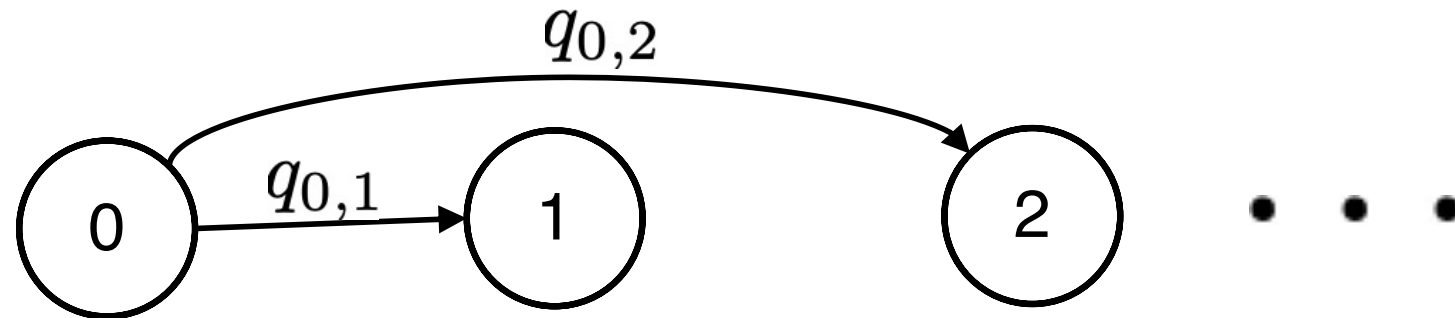
Initial condition is $P(S(i) > 0) = 1$

Solution of differential equation is

$$P(S(i) > x) = \exp(q_{i,i}x)$$

→ Sojourn time is exponentially distributed

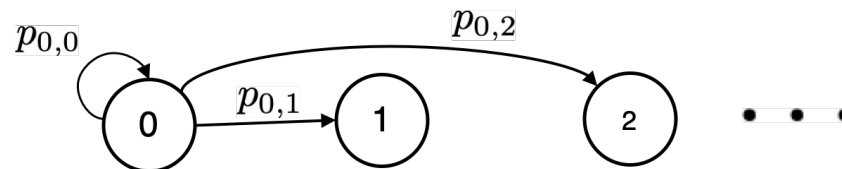
Transition Rate Diagram



- Sum of arrows out of a state i is $-q_{i,i}$

Transition Diagram

- For DTMC



Sum of arrows **out** of a state is 1

- Probability to go to state j when in i is

$$p(i, j) = \frac{q_{i,j}}{\sum_{j \neq i} q_{i,j}} = \frac{q_{i,j}}{-q_{i,i}} \Rightarrow q_{i,j} = -q_{i,i}p(i, j)$$

Transient State Distribution

- We want probability that the system is in state i at time t

$$\pi_i(t) := P(X(t) = i)$$

- Law of total probabilities: for any j

$$\pi_j(t+h) = \sum_{i \in \mathcal{E}} p_{i,j}(h) \pi_i(t) = \sum_{i \neq j} p_{i,j}(h) \pi_i(t) + p_{j,j}(h) \pi_j(t)$$

- Subtract $\pi_j(t)$ from both sides then divide by h

$$\frac{\pi_j(t+h) - \pi_j(t)}{h} = \sum_{\substack{i \in \mathcal{E} \\ i \neq j}} \frac{p_{i,j}(h)}{h} \pi_i(t) + \frac{p_{j,j}(h) - 1}{h} \pi_j(t)$$

- $h \rightarrow 0 : \frac{d\pi_j(t)}{dt} = \sum_{\substack{i \in \mathcal{E} \\ i \neq j}} q_{i,j} \pi_i(t) + q_{j,j} \pi_j(t) = \sum_{i \in \mathcal{E}} q_{i,j} \pi_i(t)$

Transient State Distribution

- For any j ,
$$\frac{d\pi_j(t)}{dt} = \sum_{i \in \mathcal{E}} q_{i,j} \pi_i(t)$$
- Row vector
$$\pi(t) = (\pi_i(t), i \in \mathcal{E})$$
- In matrix notation
$$\frac{d}{dt} \pi(t) = \pi(t) \mathbf{Q}, \quad t \geq 0$$
- Solution
$$\pi(t) = \pi(0) e^{\mathbf{Q}t}, \quad t \geq 0$$
- By definition
$$e^{\mathbf{Q}t} = \sum_{k=0}^{\infty} \frac{(\mathbf{Q}t)^k}{k!}$$
- For DTMC
$$\pi(n) = \pi(0) \mathbf{P}^n$$

Limiting State Distribution

- From equation giving transient distribution

$$\pi(t) = \pi(0) e^{\mathbf{Q}(t-h)} e^{\mathbf{Q}h} = \pi(t-h) e^{\mathbf{Q}h}$$

- For DTMC $\pi(n) = \pi(n-m) \mathbf{P}^m$

- If limit exists $\pi = \lim_{t \rightarrow \infty} \pi(t) = (\pi_i, i \in \mathcal{E})$

- $h \rightarrow 0$: $\pi(t) = \pi(t)(\mathbf{I} + \mathbf{Q}h + o(h))$

- $t \rightarrow \infty$: $\pi = \pi + \pi \mathbf{Q}h + \pi o(h)$

- Divide by h and take limit as $h \rightarrow 0$

$$\pi \mathbf{Q} = 0$$

- Normalization

$$\pi \mathbf{1} = 1$$

$$\text{For DTMC } \pi \mathbf{P} = \pi$$

CTMC Properties

- A CTMC is irreducible if all pairs of states communicate
- A CTMC is positive recurrent if all states are positive recurrent
- A CTMC is ergodic if irreducible and positive recurrent
- Irreducible + finite state-space \rightarrow positive recurrent
 \rightarrow ergodic
- \rightarrow long-run distribution = invariant measure
= limiting distribution

Existence of Limiting Distribution

- If homogeneous CTMC is **irreducible**
- If system of equations

$$\pi \mathbf{Q} = 0$$

$$\pi \mathbf{1} = 1$$

has unique strictly positive solution

$$\rightarrow \lim_{t \rightarrow \infty} \left(P(X(t) = i), i \in \mathcal{E} \right) = \pi$$

The limiting distribution exists and it is the stationary distribution

15 minutes break

Balance Equations

- Develop $\pi \mathbf{Q} = 0$

- For all $i \in \mathcal{E}$
$$\sum_{j \in \mathcal{E}} \pi_j q_{j,i} = 0$$
$$-\pi_i q_{i,i} = \sum_{j \neq i} \pi_j q_{j,i}$$

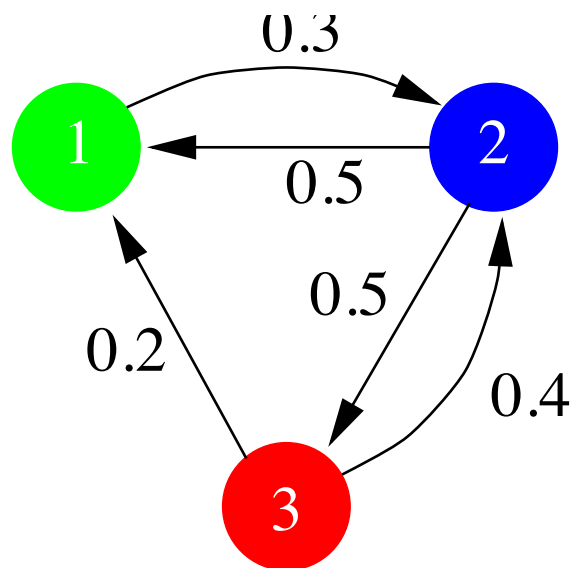
$$\left(\sum_{j \neq i} q_{i,j} \right) \pi_i = \sum_{j \neq i} q_{j,i} \pi_j$$

probability flow rate out of state i = probability flow rate into state i

- $q_{j,i}$: transition rate to state i when in state j

- $q_{j,i} \pi_j$: probability flow rate from state j to state i

Example



$$\mathcal{E} = \{1, 2, 3\}$$

CTMC irreducible? ✓

$$Q = \begin{pmatrix} -0.3 & 0.3 & 0 \\ 0.5 & -1 & 0.5 \\ 0.2 & 0.4 & -0.6 \end{pmatrix}$$

flow out = flow in

$$\begin{cases} 0.3\pi_1 & = & 0.5\pi_2 + 0.2\pi_3 \\ (0.5 + 0.5)\pi_2 & = & 0.3\pi_1 + 0.4\pi_3 \\ (0.2 + 0.4)\pi_3 & = & 0.5\pi_2 \\ \pi_1 + \pi_2 + \pi_3 & = & 1 \end{cases} \Rightarrow \begin{cases} \pi_2 = 0.3\pi_1 + 0.4\pi_3 \\ 1.2\pi_3 = \pi_2 \\ \pi_1 + \pi_2 + \pi_3 = 1 \end{cases}$$

Example

$$\left\{ \begin{array}{l} \pi_2 = 0.3\pi_1 + 0.4\pi_3 \\ 1.2\pi_3 = \pi_2 \quad \checkmark \\ \pi_1 + \pi_2 + \pi_3 = 1 \end{array} \right. \Rightarrow \begin{array}{l} 1.2\pi_3 = 0.3\pi_1 + 0.4\pi_3 \\ \Rightarrow \pi_1 = \frac{8}{3}\pi_3 \quad \checkmark \end{array}$$

■ Normalization

$$\pi_3 \left(\frac{8}{3} + \frac{6}{5} + 1 \right) = 1 \Leftrightarrow \pi_3 \left(\frac{40 + 18 + 15}{15} \right) = 1 \Leftrightarrow \pi_3 = \frac{15}{73}$$

$$\Rightarrow \pi = (\pi_1, \pi_2, \pi_3) = \frac{1}{73} (40, 18, 15) \quad \begin{array}{l} \text{sanity check} \\ \Sigma = 1 \end{array}$$

■ Solution is unique and strictly positive

→ this is the limiting distribution

Recap

	DTMC	CTMC
Chapman-Kolmogorov	$\mathbf{P}^{(n+m)} = \mathbf{P}^{(n)} \mathbf{P}^{(m)}$	$\mathbf{P}(t + s) = \mathbf{P}(t) \cdot \mathbf{P}(s)$
Main matrix	Transition matrix \mathbf{P}	Infinitesimal generator \mathbf{Q}
with characteristics	Row sum is 1	Row sum is 0
Diagram	Transition probabilities	Transition rates
with characteristics	Arrows out sum to 1	No loops
Transient distribution	$\pi(n) = \pi(0) \mathbf{P}^n$	$\pi(t) = \pi(0) e^{\mathbf{Q}t}$
Stationary distribution	$\pi \mathbf{P} = \pi \quad \pi \mathbf{1} = 1$	$\pi \mathbf{Q} = 0 \quad \pi \mathbf{1} = 1$
Properties to check	Aperiodicity, irreducibility	Irreducibility
Sojourn time	Geometric	$\text{Exp}(-q_{i,i})$
Markov property	Easy to check	????

$$P(X(t) = j \mid X(s_1) = i_1, \dots, X(s_{n-1}) = i_{n-1}, X(s) = i) \\ = P(X(t) = j \mid X(s) = i) = p_{i,j}(t - s)$$

$$i_1, \dots, i_{n-1}, i, j \in \mathcal{E}$$

$$0 \leq s_1 < \dots < s_{n-1} < s < t$$

Impossible to check
for CTMC

Construction Rule #1

- Continuous-time stochastic process $\{X(t), t \geq 0\}$
 - If for each state i
 - ▶ Process stays in i for a time that is $\text{Exp}(\tau_i)$
 - ▶ Once sojourn time is over, process jumps to state j with prob a_{ij} ($a_{ii} = 0$, their sum for all j is 1)
- $\{X(t), t \geq 0\}$ is a CTMC

- Infinitesimal generator

$$\mathbf{Q} = [q_{i,j}]_{i,j \in \mathcal{E}}, \quad q_{i,j} = \begin{cases} \tau_i a_{ij} & i \neq j \\ -\tau_i & i = j \end{cases}$$

Construction Rule #2

- Continuous-time stochastic process $\{X(t), t \geq 0\}$
 - For each state i , as soon as process enters state i
 - ▶ For each other state j generate sample for $Y_{i,j}$ from $\text{Exp}(\mu_{i,j})$
(if no transition from i to j , then $\mu_{i,j}$ is 0 and sample is ∞)
 - ▶ The first sample to expire makes process jump to corresponding state
- $\{X(t), t \geq 0\}$ is a CTMC
- Infinitesimal generator

$$\mathbf{Q} = [q_{i,j}]_{i,j \in \mathcal{E}}, \quad q_{i,j} = \begin{cases} \mu_{i,j} & j \neq i \\ -\sum_{k \neq i} \mu_{i,k} & j = i \end{cases}$$

Construction Rule #2

- In practice we use this construction when at each state i
 - ▶ Several processes or events can cause a state change
 - ▶ Each lasts for a time $\text{Exp}()$
- Example:
 - ▶ Jobs are submitted to a server according to Poisson process rate $\lambda \rightarrow$ time to submit job is $\text{Exp}(\lambda)$
 - ▶ A job has 1 task with prob $\frac{1}{2}$ and 2 tasks with prob $\frac{1}{2}$
 - ▶ Server processes tasks one at a time, service time $\text{Exp}(\mu)$
 - ▶ State is number of tasks, state-space is set of integers

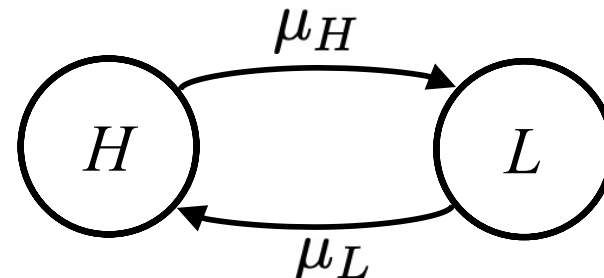
$$\left. \begin{array}{ll} i \rightarrow i + 1 & \text{Exp}(\lambda/2) \\ i \rightarrow i + 2 & \text{Exp}(\lambda/2) \\ i \rightarrow i - 1 & \text{Exp}(\mu) \quad i > 0 \end{array} \right\} \text{ we have a CTMC}$$

Example 2.1 Page 19

- Stochastic process $\mathbf{Y} = \{Y(t), t \geq 0\}$ alternates between states H and L
- Sojourn time in H is $\text{Exp}(\mu_H)$ with mean $1/\mu_H$
- Sojourn time in L is $\text{Exp}(\mu_L)$ with mean $1/\mu_L$
- $\mathcal{E} = \{H, L\}$ prob 1 to switch state
- Construction rule #1 $\rightarrow \mathbf{Y}$ is a CTMC

- Infinitesimal generator $\mathbf{Q} = \begin{pmatrix} -\mu_H & \mu_H \\ \mu_L & -\mu_L \end{pmatrix}$

- Transition diagram



Global Balance Equations

- Balance equations: for any state i

$$\left(\sum_{j \neq i} q_{i,j} \right) \pi_i = \sum_{j \neq i} q_{j,i} \pi_j$$

probability flow rate out of state i

= probability flow rate into state i

- Global balance equations: S subset of \mathcal{E}

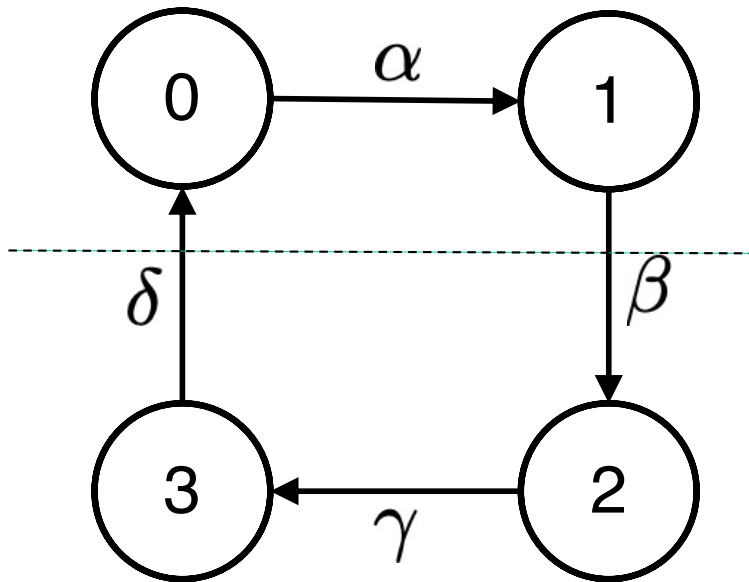
probability flow rate out of S = probability flow rate into S

$$\sum_{i \in S} \sum_{j \in \bar{S}} \pi_i q_{i,j} = \sum_{i \in \bar{S}} \sum_{j \in S} \pi_i q_{i,j}$$

► If $S = \{i\} \rightarrow$ balance equation for state i

Global Balance Equations

- Proof: use $\pi \mathbf{Q} = 0$ and $\mathbf{Q} \mathbf{1} = 0$ and find that diff is 0
- Example $\mathcal{S} = \{0, 1\}$



global balance equation

$$\beta\pi_1 = \delta\pi_3$$

balance equations

$$\alpha\pi_0 = \delta\pi_3$$

$$\beta\pi_1 = \alpha\pi_0$$

$$\gamma\pi_2 = \beta\pi_1$$

$$\delta\pi_3 = \gamma\pi_2$$

- By summing first two expressions we find the result

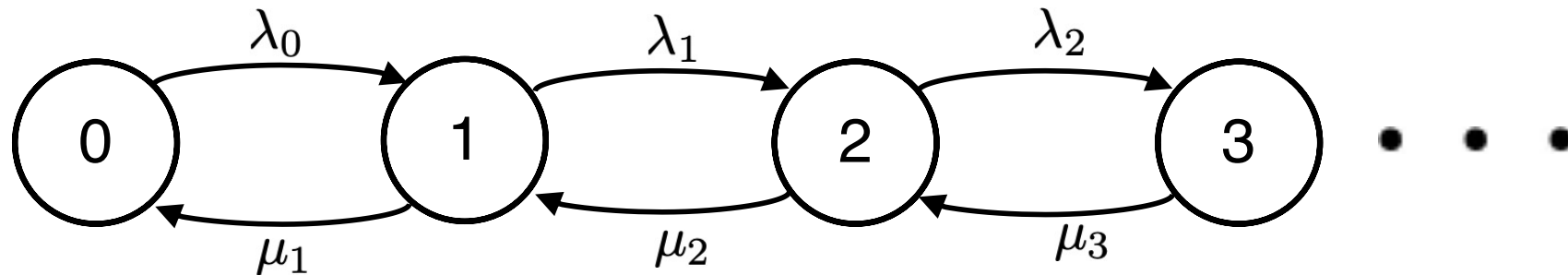
Birth and Death Process

- Particular CTMC over **set of integers** (or a subset)
 - ▶ New state is a neighbor of the previous state
- Birth and Death process seen as **population size**
- **Birth** rate when state is $i \geq 0$, $\lambda_i = q_{i,i+1}$
- **Death** rate when state is $i \geq 1$, $\mu_i = q_{i,i-1}$
- Infinitesimal generator is tri-diagonal matrix

$$Q = \begin{pmatrix} -\lambda_0 & \lambda_0 & 0 & \dots & \dots & \dots \\ \mu_1 & -(\lambda_1 + \mu_1) & \lambda_1 & 0 & \dots & \dots \\ 0 & \mu_2 & -(\lambda_2 + \mu_2) & \lambda_2 & 0 & \dots \\ \vdots & 0 & \mu_3 & -(\lambda_3 + \mu_3) & \lambda_3 & \ddots \\ & \vdots & 0 & \mu_4 & \ddots & \ddots \\ & & \vdots & & \ddots & \ddots \end{pmatrix}$$

Birth and Death Process

- Transition rate diagram $\mathcal{E} = \mathbb{N}$



- Balance equations

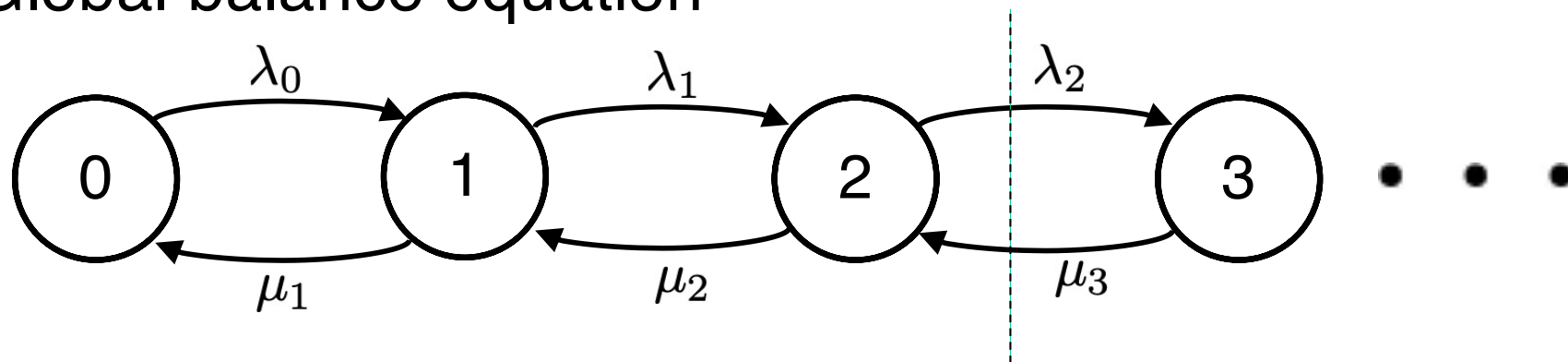
$$\begin{aligned}
 \cancel{\lambda_0} \pi_0 &= \cancel{\mu_1} \pi_1 \\
 (\cancel{\lambda_1} + \cancel{\mu_1}) \pi_1 &= \cancel{\lambda_0} \pi_0 + \cancel{\mu_2} \pi_2 \\
 &\vdots \\
 (\cancel{\lambda_{i-1}} + \cancel{\mu_{i-1}}) \pi_{i-1} &= \cancel{\lambda_{i-2}} \pi_{i-2} + \cancel{\mu_i} \pi_i \\
 (\cancel{\lambda_i} + \cancel{\mu_i}) \pi_i &= \cancel{\lambda_{i-1}} \pi_{i-1} + \cancel{\mu_{i+1}} \pi_{i+1}
 \end{aligned}$$

$$\lambda_i \pi_i = \mu_{i+1} \pi_{i+1}$$

$$\Rightarrow \pi_{i+1} = \frac{\lambda_i}{\mu_{i+1}} \pi_i, \quad i = 0, 1, \dots$$

Birth and Death Process

- Global balance equation



$$\lambda_2 \pi_2 = \mu_3 \pi_3$$

- For any state $i > 0$ $\lambda_{i-1} \pi_{i-1} = \mu_i \pi_i$

Birth and Death Process

- By recurrence

$$\pi_i = \frac{\lambda_0 \lambda_1 \cdots \lambda_{i-1}}{\mu_1 \mu_2 \cdots \mu_i} \pi_0, \quad i = 1, 2, \dots$$

- Normalization $\sum_{i=0}^{\infty} \pi_i = 1$

$$\pi_0 \left(1 + \frac{\lambda_0}{\mu_1} + \frac{\lambda_0 \lambda_1}{\mu_1 \mu_2} + \cdots + \frac{\lambda_0 \lambda_1 \cdots \lambda_{i-1}}{\mu_1 \mu_2 \cdots \mu_i} + \cdots \right) = 1$$

- Define $C := \sum_{i \geq 0} \prod_{j=1}^i \frac{\lambda_{j-1}}{\mu_j}$

- If $C < \infty$ (stability condition) then

$$\pi_0 = \frac{1}{C} > 0$$

- Irreducible \rightarrow limiting distribution exists ($C < \infty$)

Time-Reversibility

- A CTMC is time-reversible if for any two states $i, j \in \mathcal{E}$

$$\pi_i q_{i,j} = \pi_j q_{j,i}$$

- A birth and death process is time-reversible
- Consequence:
forwards chain and reverse chain are statistically identical and described by same transition diagram

- Example

Forward trajectory ...0121232101232101010...

Backward trajectory ...0101012321012321210...

Probability flow rates between any two states are same in either direction (e.g. #23 = #32)

For next week

- Lesson 2 to revise
- Homework 2 to return on Tuesday 24 September **before 9.00 am**
- Lesson 3 to read before Lecture 3