

Performance Evaluation of Networks

Sara Alouf

Ch 3 – Absorbing Markov Chains

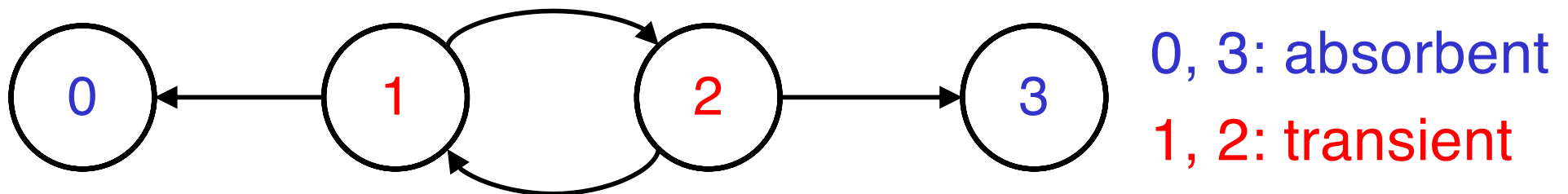
- Chapters 1 and 2: Irreducible Markov chains

- ▶ Transient distribution

- ▶ Steady-state / limiting distribution

- Absorbing Markov Chains

- some states are a dead end



- ▶ Transient distribution

- ▶ Time until absorption

- ▶ Probability to be absorbed in a given absorbing state

Discrete-Time Absorbing Markov Chain

- Homogeneous DTMC $\{X(n), n \geq 0\}$
- State space $\mathcal{E} := \{1, 2, \dots, N, 1^*, 2^*, \dots, M^*\}$
 N transient states M absorbing states
- Transition matrix $\mathbf{P} = [p_{i,j}]_{i,j \in \mathcal{E}}$

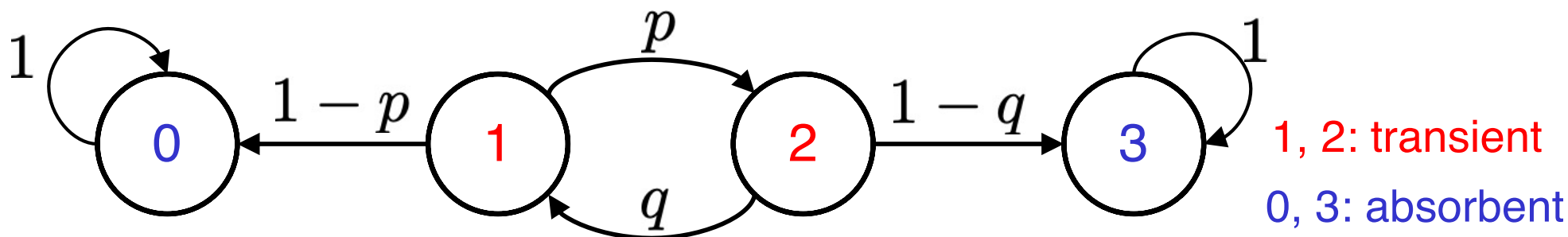
$$\mathbf{P} = \begin{pmatrix} \boxed{\begin{matrix} a_{1,1} & \dots & a_{1,N} \\ \vdots & & \vdots \\ a_{N,1} & \dots & a_{N,N} \end{matrix}} & \boxed{\begin{matrix} r_{1,1^*} & \dots & r_{1,M^*} \\ \vdots & & \vdots \\ r_{N,1^*} & \dots & r_{N,M^*} \end{matrix}} \\ \boxed{\begin{matrix} 0 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & 0 \end{matrix}} & \boxed{\begin{matrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{matrix}} \end{pmatrix} = \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$\mathbf{A} = [a_{i,j}]_{1 \leq i,j \leq N}$$

$$\mathbf{R} = [r_{i,j}]_{1 \leq i \leq N, 1^* \leq j \leq M^*}$$

Example

- DTMC with transition diagram



- State-space $\mathcal{E} = \{1, 2, 0, 3\}$ → order is important!

- Transition matrix (follow order)

$$\mathbf{A} = \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix} = \begin{bmatrix} 0 & p \\ q & 0 \end{bmatrix}$$

$$\mathbf{P} = \begin{matrix} & \begin{matrix} 1 & 2 \end{matrix} & \begin{matrix} 0 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 0 \\ 3 \end{matrix} & \begin{bmatrix} 0 & p & 1-p & 0 \\ q & 0 & 0 & 1-q \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \end{matrix} = \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$\mathbf{R} = \begin{bmatrix} r_{1,0} & r_{1,3} \\ r_{2,0} & r_{2,3} \end{bmatrix} = \begin{bmatrix} 1-p & 0 \\ 0 & 1-q \end{bmatrix}$$

n -step Transition Matrix

- Transition matrix $\mathbf{P} = \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$

- From Chapman-Kolmogorov equation $\mathbf{P}^{(n)} = \mathbf{P}^n$

$$\mathbf{P}^2 = \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix} = \begin{bmatrix} \mathbf{A}^2 & \mathbf{AR} + \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$\mathbf{P}^3 = \begin{bmatrix} \mathbf{A}^2 & \mathbf{AR} + \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix} = \begin{bmatrix} \mathbf{A}^3 & \mathbf{A}^2\mathbf{R} + \mathbf{AR} + \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

$$\mathbf{P}^n = \begin{bmatrix} \mathbf{A}^n & \sum_{k=0}^{n-1} \mathbf{A}^k \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

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

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

Fundamental Matrix

- Define $N \times N$ matrix $\mathbf{N} = [n_{i,j}]_{1 \leq i,j \leq N}$

$n_{i,j}$ expected number of visits to state j if initially in i

- Define $X_j^{(n)} = \mathbb{1}(X(n) = j) = \begin{cases} 1 & X(n) = j \\ 0 & \text{otherwise} \end{cases}$

- We have $n_{i,j} = E \left[\sum_{n \geq 0} X_j^{(n)} \mid X(0) = i \right]$

- Proposition 8: the fundamental matrix is

$$\mathbf{N} = (\mathbf{I} - \mathbf{A})^{-1}$$

Nota bene:
elements in \mathbf{N} are
positive or **null**

Proof of $\mathbf{N} = (\mathbf{I} - \mathbf{A})^{-1}$

$$\begin{aligned}
 E \left[\sum_{k=0}^n X_j^{(k)} \mid X(0) = i \right] &= \sum_{k=0}^n E \left[X_j^{(k)} \mid X(0) = i \right] \\
 &= \sum_{k=0}^n E [\mathbb{1}(X(k) = j) \mid X(0) = i] \\
 &= \sum_{k=0}^n P(X(k) = j \mid X(0) = i) \\
 &= \sum_{k=0}^n a_{i,j}^{(k)} \quad (i, j) \text{ element in } \mathbf{A}^k
 \end{aligned}$$

$$\begin{aligned}
 E \left[\lim_{n \rightarrow \infty} \sum_{k=0}^n X_j^{(k)} \mid X(0) = i \right] &= \lim_{n \rightarrow \infty} E \left[\sum_{k=0}^n X_j^{(k)} \mid X(0) = i \right] \\
 n_{i,j} &= \sum_{k \geq 0} a_{i,j}^{(k)}
 \end{aligned}$$

expected number of visits
in $n + 1$ steps

Proof of $\mathbf{N} = (\mathbf{I} - \mathbf{A})^{-1}$

- For any $i, j = 1, \dots, N$ $n_{i,j} = \sum_{k \geq 0} a_{i,j}^{(k)}$

$$\Rightarrow \boxed{\mathbf{N} = \sum_{k \geq 0} \mathbf{A}^k}$$

- Consider finite sum

$$(\mathbf{I} - \mathbf{A}) \sum_{k=0}^n \mathbf{A}^k = \sum_{k=0}^n \mathbf{A}^k - \sum_{k=1}^{n+1} \mathbf{A}^k = \mathbf{I} - \mathbf{A}^{n+1}$$

- Let $n \rightarrow \infty$ $a_{i,j}^{(n+1)} \rightarrow 0 \Rightarrow \mathbf{A}^{n+1} \rightarrow 0$

$$\rightarrow (\mathbf{I} - \mathbf{A})\mathbf{N} = \mathbf{I}$$

- If inverse exists $\rightarrow \boxed{\mathbf{N} = (\mathbf{I} - \mathbf{A})^{-1}}$

Absorption Probabilities

- Define $N \times M^*$ **stochastic** matrix $\mathbf{B} = [b_{i,j}]_{\substack{i \in \{1, \dots, N\} \\ j \in \{1^*, \dots, M^*\}}}$

$b_{i,j}$ probability to be absorbed in j if initially in i

- Proposition 9: $\mathbf{B} = \mathbf{NR}$ recall $\mathbf{P} = \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$

- For $i \in \{1, \dots, N\}, j \in \{1^*, \dots, M^*\}$

$$b_{i,j} = \sum_{n \geq 0} P(X(n+1) = j \mid X(0) = i) \quad \begin{array}{l} \text{use law of tot. prob.} \\ k \text{ last transient state} \end{array}$$

$$= \sum_{n \geq 0} \sum_{k=1}^N \overset{r_{k,j}}{P(X(n+1) = j \mid X(n) = k, \cancel{X(0) = i})}$$

$$\times \underset{a_{i,k}^{(n)}}{P(X(n) = k \mid X(0) = i)}$$

use Markov property

Absorption Probabilities

- For $i \in \{1, \dots, N\}$, $j \in \{1^*, \dots, M^*\}$

$$\begin{aligned} b_{i,j} &= \sum_{n \geq 0} \sum_{k=1}^N r_{k,j} a_{i,k}^{(n)} \\ &= \sum_{k=1}^N \left(\sum_{n \geq 0} a_{i,k}^{(n)} \right) r_{k,j} \\ &= \sum_{k=1}^N n_{i,k} r_{k,j} \end{aligned}$$

- In matrix notation

$$\mathbf{B} = \mathbf{N}\mathbf{R} = (\mathbf{I} - \mathbf{A})^{-1}\mathbf{R}$$

Nota bene:
elements in \mathbf{B} are
probabilities
 \mathbf{B} is **stochastic** matrix

Limit of \mathbf{P}^n

- n -step transition matrix

$$\mathbf{P}^n = \begin{bmatrix} \mathbf{A}^n & \sum_{k=0}^{n-1} \mathbf{A}^k \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

- As $n \rightarrow \infty$ $\mathbf{A}^n \rightarrow 0$

$$\sum_{k=0}^{n-1} \mathbf{A}^k \mathbf{R} \rightarrow \mathbf{N} \mathbf{R}$$

$$\mathbf{P}^n \rightarrow \begin{bmatrix} \mathbf{0} & \mathbf{B} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

stochastic matrix

Expected Absorption Time

- Define **column** vector $\mathbf{T} = (T(i), i = 1, \dots, N)^T$

$T(i)$ expected time until absorption if initially in state i

time = number of steps

- For any absorbing state $j \in \{1^*, \dots, M^*\}$, $T(j) = 0$

- Proposition 10: $\mathbf{T} = \mathbf{N} \cdot \mathbf{1}$

- $n_{i,j}$ expected number of visits to state j if initially in i

$$\rightarrow T(i) = \sum_{j=1}^N n_{i,j} \quad \text{for } i \in \{1, 2, \dots, N\}$$

Expected Absorption Time

- Corollary 1: **Column** vector $\mathbf{T} = (T(i), i = 1, \dots, N)^T$ is solution of $\mathbf{T} = \mathbf{1} + \mathbf{AT}$
- Fundamental matrix $\mathbf{N} = (\mathbf{I} - \mathbf{A})^{-1}$
- $\mathbf{T} = \mathbf{N}.\mathbf{1} \Leftrightarrow \mathbf{T} = (\mathbf{I} - \mathbf{A})^{-1}.\mathbf{1}$
 - $\Leftrightarrow (\mathbf{I} - \mathbf{A})\mathbf{T} = \mathbf{1}$
 - $\Leftrightarrow \mathbf{T} - \mathbf{AT} = \mathbf{1}$
 - $\Leftrightarrow \mathbf{T} = \mathbf{1} + \mathbf{AT}$

Nota bene:
elements in \mathbf{T} are
strictly positive

Example

- Document in P2P system is replicated over K peers
- Retrieve requests occur at beginning of every minute
- $X(n)$ number of replicas available just before minute n
- Peers connect/disconnect from system $\rightarrow X(n)$ stochastic
- If at n no copy is available \rightarrow request fails $X(n) = F$
- State-space $\mathcal{E} = \{1, \dots, K, F\}$
- Transition matrix

$$\mathbf{P} = \left[\begin{array}{cccc|c} \frac{1}{K+1} & \frac{1}{K+1} & \cdots & \frac{1}{K+1} & \frac{1}{K+1} \\ \vdots & \vdots & & \vdots & \vdots \\ \frac{1}{K+1} & \frac{1}{K+1} & \cdots & \frac{1}{K+1} & \frac{1}{K+1} \\ \hline 0 & 0 & \cdots & 0 & 1 \end{array} \right]$$

Example

- Describe $\{X(n), n \geq 0\}$
 - ▶ Process observed every minute when requests arrive
 - discrete-time process
 - ▶ Transition probabilities independent of step
 - homogeneous
 - ▶ States 1 to K are transient
 - ▶ State F is absorbent
- $\{X(n), n \geq 0\}$ is absorbing homogeneous DTMC
- Limiting probability of a failure?
 - ▶ Regardless of initial distribution the chain will be absorbed in its unique absorbing state
 - limiting probability of failure is 1

Example

- Compute expected absorption times when $K = 3$

$$\mathbf{P} = \begin{bmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \Rightarrow \mathbf{A} = \frac{1}{4} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

► $T(F) = 0$

► $\mathbf{T} = \mathbf{1} + \mathbf{A}\mathbf{T} \Rightarrow \begin{bmatrix} T(1) \\ T(2) \\ T(3) \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} + \frac{1}{4} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \cdot \begin{bmatrix} T(1) \\ T(2) \\ T(3) \end{bmatrix}$

$$\Rightarrow T(1) = T(2) = T(3) = 1 + \frac{1}{4}(T(1) + T(2) + T(3))$$

$$\Rightarrow 4T(1) = 4 + 3T(1) \Rightarrow T(1) = T(2) = T(3) = 4$$

► $\mathbf{N} = (\mathbf{I} - \mathbf{A})^{-1} = \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix} \quad \begin{array}{l} \sum = T(1) = 4 \\ \sum = T(2) = 4 \\ \sum = T(3) = 4 \end{array} \quad \mathbf{T} = \mathbf{N} \cdot \mathbf{1}$

15 minutes break

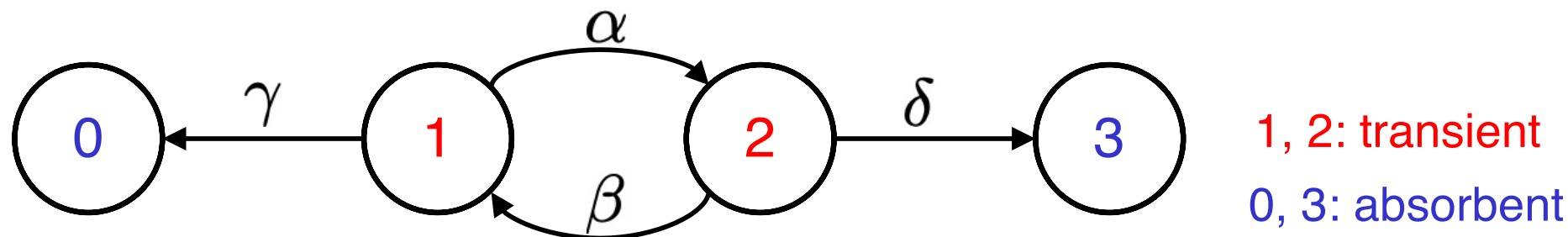
Continuous-Time Absorbing Markov Chain

- Homogeneous CTMC $\{X(t), t \geq 0\}$
- State space $\mathcal{E} := \{1, 2, \dots, N, 1^*, 2^*, \dots, M^*\}$
 N transient states M absorbing states

- Infinitesimal generator $\mathbf{Q} = [q_{i,j}]_{i,j \in \mathcal{E}}$
- $$\mathbf{Q} = \begin{pmatrix} \boxed{\begin{matrix} q_{1,1} & \dots & q_{1,N} \\ \vdots & & \vdots \\ q_{N,1} & \dots & q_{N,N} \end{matrix}} & \boxed{\begin{matrix} \tilde{r}_{1,1^*} & \dots & \tilde{r}_{1,M^*} \\ \vdots & & \vdots \\ \tilde{r}_{N,1^*} & \dots & \tilde{r}_{N,M^*} \end{matrix}} \\ \boxed{\begin{matrix} 0 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & 0 \end{matrix}} & \boxed{\begin{matrix} 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \end{matrix}} \end{pmatrix} = \begin{bmatrix} \tilde{\mathbf{Q}} & \tilde{\mathbf{R}} \\ \mathbf{0} & \mathbf{0} \end{bmatrix}$$
- $$\tilde{\mathbf{Q}} = [q_{i,j}]_{1 \leq i,j \leq N} \quad \tilde{\mathbf{R}} = [\tilde{r}_{i,j}]_{1 \leq i \leq N, 1^* \leq j \leq M^*}$$

Example

- CTMC with transition rate diagram



- State-space $\mathcal{E} = \{1, 2, 0, 3\}$ → order is important!

- Infinitesimal generator (follow order)

$$\mathbf{Q} = \begin{matrix} & \begin{matrix} 1 & 2 & 0 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 0 \\ 3 \end{matrix} & \begin{bmatrix} -(\alpha + \gamma) & \alpha & \gamma & 0 \\ \beta & -(\beta + \delta) & 0 & \delta \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \end{matrix} \\
 = \begin{bmatrix} \tilde{\mathbf{Q}} & \tilde{\mathbf{R}} \\ \mathbf{0} & \mathbf{0} \end{bmatrix}$$

$$\tilde{\mathbf{Q}} = \begin{bmatrix} -(\alpha + \gamma) & \alpha \\ \beta & -(\beta + \delta) \end{bmatrix}$$

$$\tilde{\mathbf{R}} = \begin{bmatrix} \tilde{r}_{1,0} & \tilde{r}_{1,3} \\ \tilde{r}_{2,0} & \tilde{r}_{2,3} \end{bmatrix} = \begin{bmatrix} \gamma & 0 \\ 0 & \delta \end{bmatrix}$$

Embedded Markov Chain at Jump Times

- Construct DTMC from CTMC
- Observe $\{X(t), t \geq 0\}$ at **jump** times $\rightarrow \{X(n), n \geq 0\}$
- Transition probabilities at jump times

$$p(i, j) = \begin{cases} \frac{q_{i,j}}{-q_{i,i}} & i \in \{1, \dots, N\}, j \in \mathcal{E}, j \neq i \\ 1 & j = i \in \{1^*, \dots, M^*\} \text{ (by convention)} \\ 0 & \text{otherwise} \end{cases}$$

- Transition matrix (stochastic) $\mathbf{P} = [p(i, j)]_{i,j \in \mathcal{E}} = \begin{bmatrix} \mathbf{A} & \mathbf{R} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$

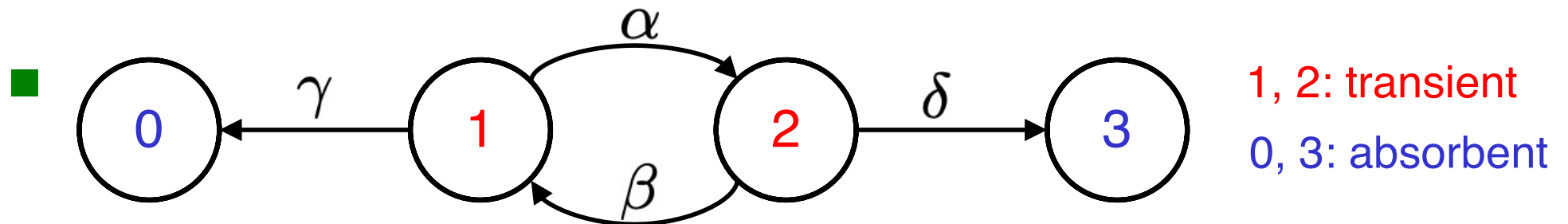
$$\mathbf{A} = [a_{i,j}]_{1 \leq i,j \leq N}$$

$$a_{i,j} = \begin{cases} \frac{q_{i,j}}{-q_{i,i}} & j \neq i \\ 0 & j = i \end{cases}$$

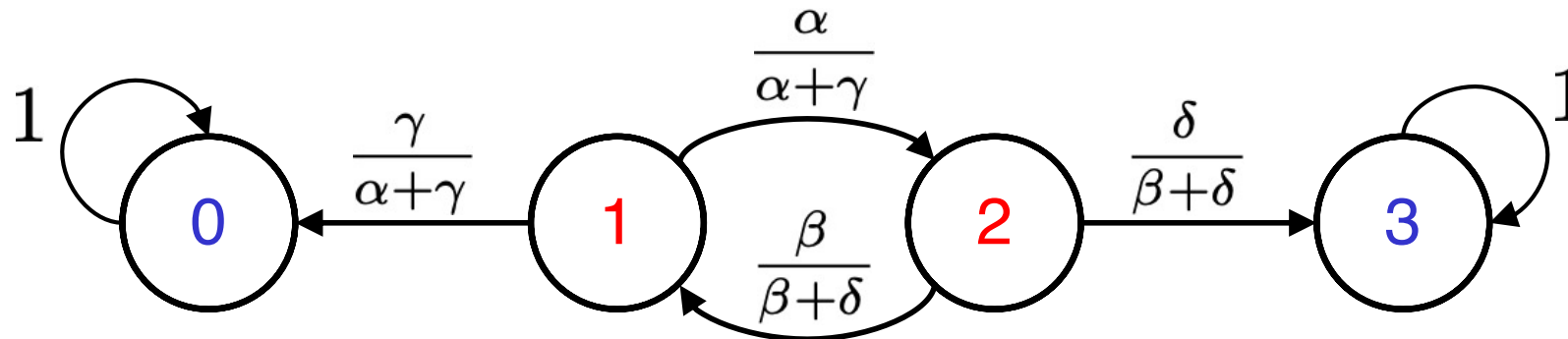
$$\mathbf{R} = [r_{i,j}]_{1 \leq i \leq N, 1^* \leq j \leq M^*}$$

$$r_{i,j} = \frac{q_{i,j}}{-q_{i,i}}$$

Example



■ Embedded Markov chain at jump times



$$\mathbf{P} = \begin{array}{c} \begin{array}{cc} 1 & 2 \\ \begin{array}{c} 0 \\ 2 \end{array} & \begin{array}{c} \frac{\alpha}{\alpha+\gamma} \\ 0 \end{array} \\ \begin{array}{c} 0 \\ 3 \end{array} & \begin{array}{c} \frac{\gamma}{\alpha+\gamma} \\ 0 \end{array} \end{array} \left| \begin{array}{cc} 0 & 3 \\ \begin{array}{c} \frac{\gamma}{\alpha+\gamma} \\ 0 \end{array} & \begin{array}{c} 0 \\ 1 \end{array} \end{array} \right] \end{array}$$

$$\mathbf{A} = \begin{bmatrix} 0 & \frac{\alpha}{\alpha+\gamma} \\ \frac{\beta}{\beta+\delta} & 0 \end{bmatrix}$$

$$\mathbf{R} = \begin{bmatrix} \frac{\gamma}{\alpha+\gamma} & 0 \\ 0 & \frac{\delta}{\beta+\delta} \end{bmatrix}$$

Mean Number of Visits

- Use embedded Markov chain and results of absorbing DTMC

$$\mathbf{N} = [n_{i,j}]_{1 \leq i,j \leq N} \quad \boxed{\mathbf{N} = (\mathbf{I} - \mathbf{A})^{-1}}$$

$n_{i,j}$ expected number of visits to state j if initially in i

- Visits durations are different!
- In absorbing DTMC
 - a visit lasts for a constant step time
- In Markov chain embedded at jump times of absorbing CTMC
 - a visit in transient i lasts for a random time that is $\text{Exp}(-q_{i,i})$

Absorption Probabilities

- Use embedded Markov chain and results of absorbing DTMC

$$\mathbf{B} = \left[b_{i,j} \right]_{\substack{i \in \{1, \dots, N\} \\ j \in \{1^*, \dots, M^*\}}} \quad \mathbf{B} = \mathbf{N}\mathbf{R} = (\mathbf{I} - \mathbf{A})^{-1}\mathbf{R}$$

$b_{i,j}$ probability to be absorbed in j if initially in i

- Matrix is stochastic

Expected Absorption Time

- In absorbing CTMC

number of visits to transient state \neq expected time spent in that state

- Define **column** vector $\mathbf{T} = (T(i), i = 1, \dots, N)^T$

$T(i)$ expected time until absorption if initially in state i

time \neq number of steps

- Define $N \times N$ matrix $\tilde{\mathbf{T}} = [t_{i,j}]_{1 \leq i,j \leq N}$

$t_{i,j}$ expected time spent in transient j if initially in state i

- $T(i) = \sum_{j=1}^N t_{i,j}$ $t_{i,j} = n_{i,j} \times \frac{1}{-q_{j,j}}, \quad i, j = 1, \dots, N$

Expected Absorption Time

■ Proposition 11: $\tilde{\mathbf{T}} = -\tilde{\mathbf{Q}}^{-1}$ $\mathbf{T} = -\tilde{\mathbf{Q}}^{-1} \cdot \mathbf{1}$

■ Define $\mathbf{D} = \text{diag} \left(\frac{1}{q_{1,1}}, \dots, \frac{1}{q_{N,N}} \right)$

$$\mathbf{ND} = \begin{bmatrix} n_{1,1} & \dots & n_{1,N} \\ \vdots & & \vdots \\ n_{N,1} & \dots & n_{N,N} \end{bmatrix} \begin{bmatrix} \frac{1}{q_{1,1}} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{q_{N,N}} \end{bmatrix} = \begin{bmatrix} \frac{n_{1,1}}{q_{1,1}} & \dots & \frac{n_{1,N}}{q_{N,N}} \\ \vdots & & \vdots \\ \frac{n_{N,1}}{q_{1,1}} & \dots & \frac{n_{N,N}}{q_{N,N}} \end{bmatrix}$$

$$\Rightarrow \tilde{\mathbf{T}} = -\mathbf{ND} = -(\mathbf{I} - \mathbf{A})^{-1} \mathbf{D} = -(\mathbf{D}^{-1} (\mathbf{I} - \mathbf{A}))^{-1}$$

$$\begin{aligned} \mathbf{D}^{-1} (\mathbf{I} - \mathbf{A}) &= \begin{bmatrix} q_{1,1} & & 0 \\ & \ddots & \\ 0 & & q_{N,N} \end{bmatrix} \begin{bmatrix} 1 & \dots & \frac{q_{1,N}}{q_{1,1}} \\ \vdots & & \vdots \\ \frac{q_{N,1}}{q_{N,N}} & \dots & 1 \end{bmatrix} \\ &= \begin{bmatrix} q_{1,1} & \dots & q_{1,N} \\ \vdots & & \vdots \\ q_{N,1} & \dots & q_{N,N} \end{bmatrix} = \tilde{\mathbf{Q}} \end{aligned}$$

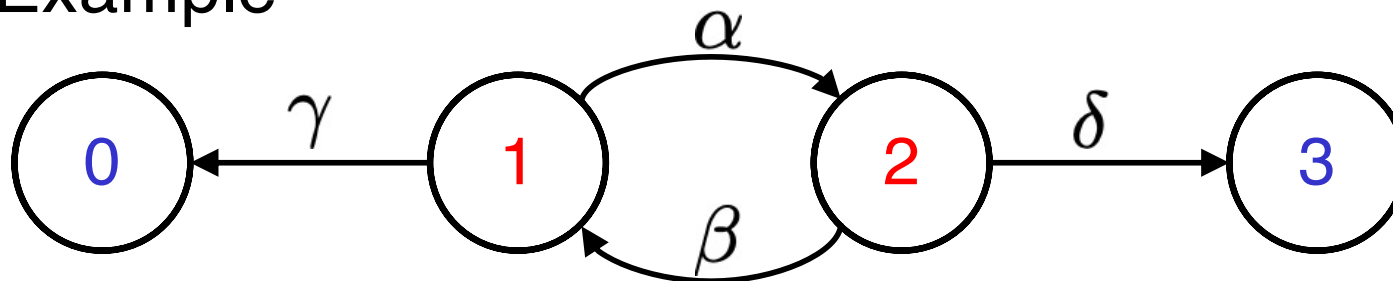
Nota bene:
elements in \mathbf{T} are
strictly positive

$$\Rightarrow \tilde{\mathbf{T}} = -\tilde{\mathbf{Q}}^{-1}$$

Expected Absorption Time

- Corollary 2: **Column** vector $\mathbf{T} = (T(i), i = 1, \dots, N)^T$ is solution of $\tilde{\mathbf{Q}}\mathbf{T} = -\mathbf{1}$

- Example



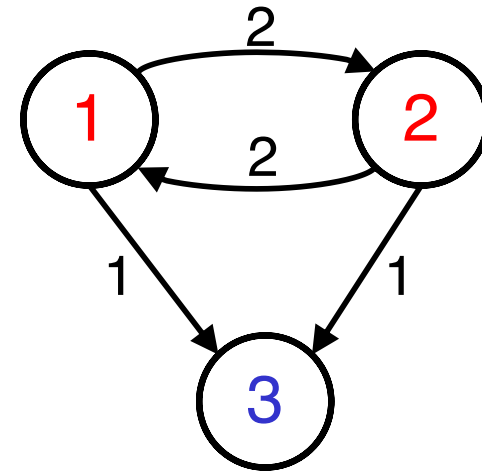
$$\tilde{\mathbf{Q}}\mathbf{T} = \begin{bmatrix} -(\alpha + \gamma) & \alpha \\ \beta & -(\beta + \delta) \end{bmatrix} \begin{bmatrix} T(1) \\ T(2) \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

$$\Rightarrow \begin{cases} -(\alpha + \gamma) T(1) + \alpha T(2) = -1 \\ \beta T(1) - (\beta + \delta) T(2) = -1 \end{cases} \Rightarrow \begin{cases} T(1) = \frac{\alpha + \beta + \delta}{\beta\gamma + \delta\alpha + \delta\gamma} \\ T(2) = \frac{\alpha + \beta + \gamma}{\beta\gamma + \delta\alpha + \delta\gamma} \end{cases}$$

Example 4 page 30

- Absorbing homogeneous CTMC

$$\mathbf{Q} = \left[\begin{array}{cc|c} -3 & 2 & 1 \\ 2 & -3 & 1 \\ \hline 0 & 0 & 0 \end{array} \right]$$



- Expected absorption time

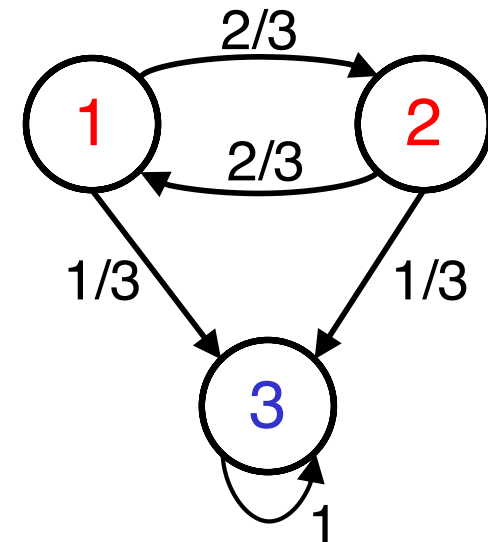
$$-3T(1) + 2T(2) = -1$$

$$2T(1) - 3T(2) = -1$$

$$\Rightarrow T(1) = T(2) = 1$$

- Embedded Markov chain

$$\mathbf{P} = \left[\begin{array}{cc|c} 0 & \frac{2}{3} & \frac{1}{3} \\ \frac{2}{3} & 0 & \frac{1}{3} \\ \hline 0 & 0 & 1 \end{array} \right] \Rightarrow \mathbf{A} = \left[\begin{array}{cc} 0 & \frac{2}{3} \\ \frac{2}{3} & 0 \end{array} \right] \quad \mathbf{R} = \left[\begin{array}{c} \frac{1}{3} \\ \frac{1}{3} \end{array} \right]$$



Example 4 page 30

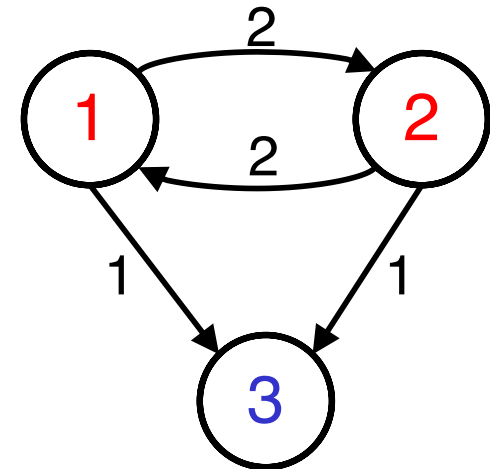
- Mean number of visits in transient states

$$\mathbf{I} - \mathbf{A} = \begin{bmatrix} 1 & -\frac{2}{3} \\ -\frac{2}{3} & 1 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 3 & -2 \\ -2 & 3 \end{bmatrix}$$

$$(\mathbf{I} - \mathbf{A})^{-1} = 3 \begin{bmatrix} 3 & -2 \\ -2 & 3 \end{bmatrix}^{-1} = \frac{3}{5} \begin{bmatrix} 3 & 2 \\ 2 & 3 \end{bmatrix} = \mathbf{N}$$

- Absorbing probabilities

$$\mathbf{B} = \mathbf{NR} = \frac{3}{5} \begin{bmatrix} 3 & 2 \\ 2 & 3 \end{bmatrix} \frac{1}{3} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} b_{1,3} \\ b_{2,3} \end{bmatrix}$$



For next week

- Lesson 3 to revise
- Homework 3 to return on Tuesday 1 October before 9 am
- Lesson 4 to read before Lecture 4