

Time reversibility

Let $X(t)$ be an irreducible CTMC with stationary distribution π . Assume that $X(0) \sim \pi$, so that for all $t > 0$, $X(t) \sim \pi$ as well.

The reversed process: Fix a large t , and consider the reverse chain $\bar{X}(s) := X(t-s)$, for $0 \leq s \leq t$.

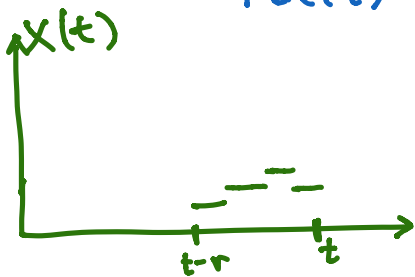
Claim: conditional on $\bar{X}(0) = i$, the time \bar{X} spends in state i is $\text{Exp}(\nu_i)$, just like X .

Proof:
$$P(\bar{T}_i \geq r) = P(\bar{X}(s) = i \text{ for } s \in [0, r] \mid \bar{X}(0) = i)$$

$$= P(X(t-s) = i \text{ for } s \in [0, r] \mid X(t) = i)$$

$$= \frac{P(X(t-s) = i \text{ for } s \in [0, r])}{P(X(t) = i)}$$

$$= \frac{P(X(t-s) = i \text{ for } s \in [0, r] \mid X(t-r) = i) P(X(t-r) = i)}{P(X(t) = i)}$$



$$= \frac{P(T_i \geq r)}{\cancel{\pi_i}} \cdot \cancel{\pi_i}$$

$$= e^{-\nu_i r} \Rightarrow \bar{T}_i \sim \text{Exp}(\nu_i)$$

time that \bar{X} spends in state i .

Note that the embedded chain of \bar{X} is the time-reversal of the embedded chain of X .

\Rightarrow By our results for discrete-time MC's, the embedded chain of \bar{X} has transition probabilities

$\bar{p}_{ij} = \frac{\sigma_j p_{ji}}{\sigma_i}$, where σ is the stationary distribution of the embedded chain of X .

Recall that last week we showed that

$$\pi_i = \frac{\sigma_i}{\nu_i} \cdot \frac{1}{Z} \leadsto \text{normalizing constant.}$$

$$\Rightarrow \sigma_i = \pi_i \nu_i Z$$

\Rightarrow The embedded chain of \bar{X} has transitions:

$$\bar{p}_{ij} = \frac{\pi_j \nu_j \cancel{Z} p_{ji}}{\pi_i \nu_i \cancel{Z}} = \frac{\pi_j \nu_j p_{ji}}{\pi_i \nu_i}.$$

Therefore, \bar{X} is the same as X if and only if

$$\bar{p}_{ij} = p_{ij}, \text{ i.e., } \frac{\pi_j \nu_j p_{ji}}{\pi_i \nu_i} = p_{ij}$$

$$\Leftrightarrow \pi_j \underbrace{\nu_j p_{ji}}_{q_{ji}} = \pi_i \underbrace{\nu_i p_{ij}}_{q_{ij}}$$

$$\Leftrightarrow \pi_j q_{ji} = \pi_i q_{ij}.$$

Recall:
 $\pi Q = \pi \cdot$

Thus we get:

Definition: The CTMC X is **time-reversible** if it has a stationary distribution π and $\pi_i q_{ij} = \pi_j q_{ji} \forall i, j$.

Like in the discrete case, $LHS = \pi_i q_{ij} = \pi_i \underbrace{\nu_i}_{\text{prob. of being in } i} \underbrace{p_{ij}}_{\text{freq. of leaving } i}$ can be interpreted as the frequency of jumping from i to j and $RHS = \pi_j q_{ji} = \pi_j \nu_j p_{ji}$ as the frequency of jumping from j to i .

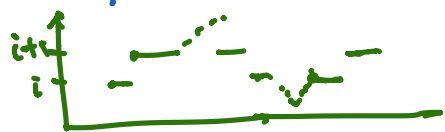
Example: birth and death processes.

Proposition: Any birth and death process with a stationary distribution is time-reversible.

Proof idea: Between 2 consecutive transitions from i to $i+1$

$i \rightarrow i+1 \rightarrow \dots \rightarrow i \rightarrow i+1$

The process has to go from $i+1$ to i exactly once.



\Rightarrow In the long run #jumps from i to $i+1$ will equal #jumps from $i+1$ to i .

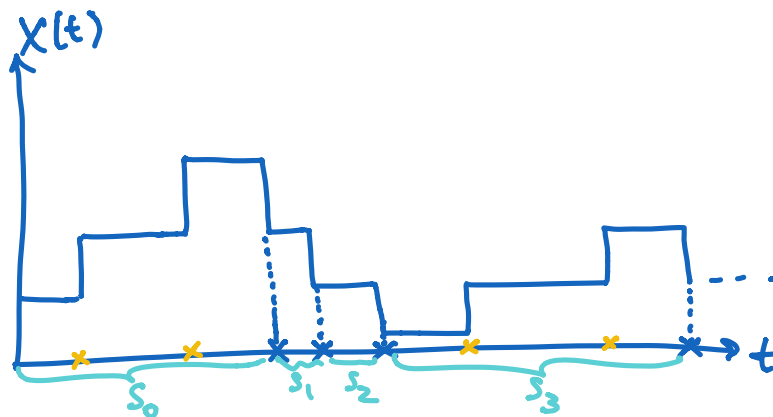
Moreover, for $|i-j| > 1$, both sides of $\pi_i q_{ij} = \pi_j q_{ji}$ are 0.

The M/M/s queue: s servers; new customers arrive at a rate- λ Poisson process. When n customers are present, they leave at rate

$$\mu_n = \begin{cases} n\mu & \text{if } n \leq s \\ s\mu & \text{if } n > s \end{cases} \quad \text{all servers are busy}$$

Corollary: For the M/M/s queue, if $\lambda < s\mu$, the departures of the clients form a Poisson process of rate λ .

Proof:



First, we show a stationary distribution exists.

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

$$\begin{aligned} \mu_1 \dots \mu_i &= \mu_1 \dots \mu_s \mu_{s+1} \dots \mu_i \\ &= \mu^s s! \mu^{i-s} s^{i-s} \\ &= \mu^i s! s^{i-s} \end{aligned}$$

In the case of M/M/s queue:

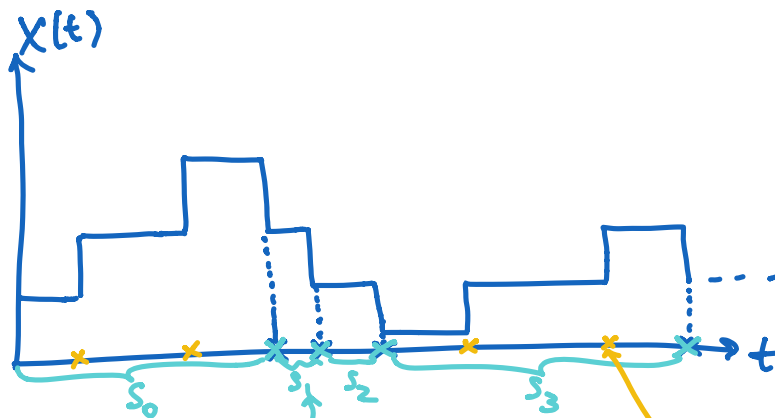
$$r_i = \begin{cases} \frac{\lambda^i}{i! \mu^i} & , i \leq s \\ \frac{\lambda^i}{s! s^{i-s} \mu^i s^s} & , i \geq s \end{cases}$$

$$r_i = \begin{cases} \left(\frac{\lambda}{\mu}\right)^i \frac{1}{i!} & , i \leq s \\ \left(\frac{\lambda}{s\mu}\right)^i \cdot \frac{s^s}{s!} & , i \geq s \end{cases}$$

The stationary distribution π exists if $\sum_{i=1}^{\infty} r_i < \infty$.

$$\begin{aligned} \sum_{i=1}^{\infty} r_i &= \underbrace{\sum_{i=1}^s \left(\frac{\lambda}{\mu}\right)^i \frac{1}{i!}}_{\text{const}} + \underbrace{\left(\sum_{i=s+1}^{\infty} \left(\frac{\lambda}{s\mu}\right)^i\right) \frac{s^s}{s!}}_{= \left(\frac{\lambda}{s\mu}\right)^{s+1} \sum_{i=0}^{\infty} \left(\frac{\lambda}{s\mu}\right)^i} < \infty \\ &\Leftrightarrow \lambda < s\mu. \end{aligned}$$

Thus, stationary distribution π exists $\Leftrightarrow \lambda < s\mu$.



By the Proposition, $X(t)$ is time-reversible. Therefore, the reverse process $\bar{X}(t)$ has the same parameters λ_i and p_{ij} as $X(t)$.

Recall that the times $X(t)$ increases by 1 form a Poisson(λ) process.

\Rightarrow the times $\bar{X}(t)$ increases by 1 form a Poisson(λ) process.

But those are the same as the times $X(t)$

decreases by 1. \Rightarrow The client departures also form a Poisson(λ) process.