

$\{X_n\}$ MC, transition prob (P_{ij}) , with stat dist π ,

$Y_n \equiv X_{N-n}$ is MC with $P(Y_{n+1}=j|Y_n=i) = \frac{\pi_j}{\pi_i} P_{ji}$.

$\{X_n\}$ reversible if $\{X\}$, $\{Y\}$ have same dynamics.



Detailed Balance Condition

$$\pi_i P_{ij} = \pi_j P_{ji} \quad (\forall i, j)$$

Thm: μ satisfies DBC, then μ is stat meas.

pf: $\sum_i \mu_i P_{ij} = \mu_j \sum_i P_{ji} = \mu_j$.

Only sufficient, not necessary.

DBC important for Markov Chain Monte Carlo (MCMC) methods, e.g., Metropolis-Hastings

e.g.: $P_{i,i+1} = 1 - \frac{i}{m}$, $P_{i,i-1} = \frac{i}{m}$ for $i \in \{0, 1, \dots, m\}$,

DBC: $\mu_i \cdot (1 - \frac{i}{m}) = \mu_{i+1} \frac{i+1}{m}$, $\frac{\mu_{i+1}}{\mu_i} = \frac{m-i}{i+1}$

$$\frac{\mu_i}{\mu_0} = \frac{m!}{i!(m-i)!} = \binom{m}{i}, \text{ set } \mu_0 = 1, \mu_i = \binom{m}{i}$$

and $\sum_{i=1}^m \mu_i = 2^m$, so stat dist $\boxed{\pi_i = \frac{1}{2^m} \cdot \binom{m}{i}}$

e.g.: G finite connected graph, no loops and multiple edges, check DBC holds in equilibrium for RW on graph.

$$\text{Def: } \pi_i = \frac{d_i}{\sum_j d_j} \quad \text{and} \quad p_{ij} = \frac{1}{d_i}$$

$$\text{so } \pi_i \cdot p_{ij} = \frac{1}{\sum_j d_j} = \pi_j \cdot p_{ji} \quad \checkmark$$

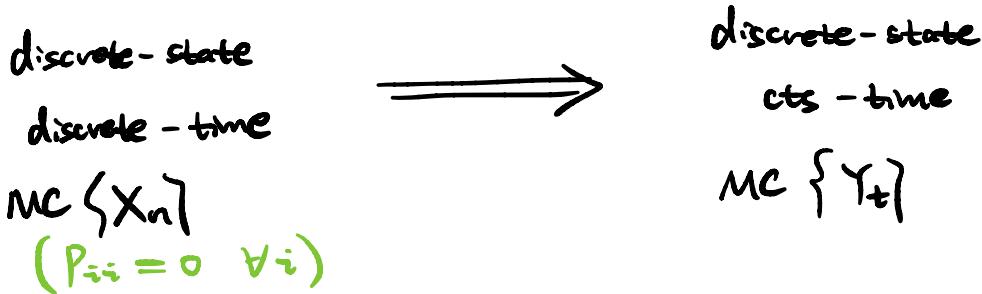
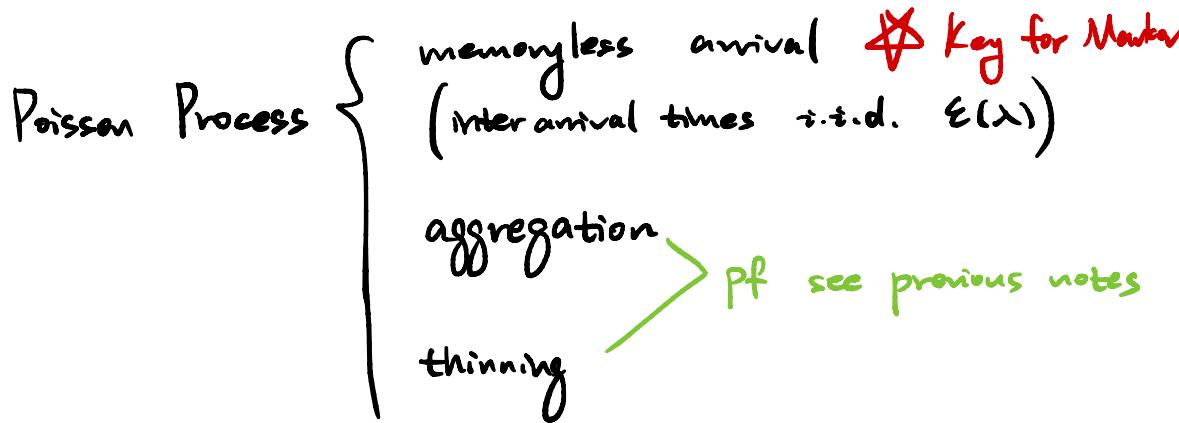
Rem: along each undirected edge  the symmetry does not hold when $d_i \neq d_j$.
 If $p_{ij} = f(i, j)$, then $p_{ji} = f(j, i)$

If impose symmetry, together with DBC for start dist,

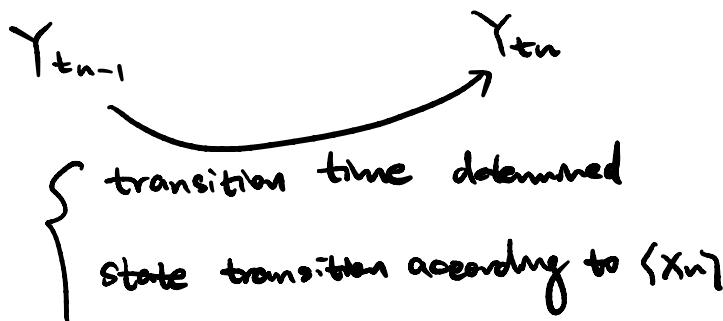
$$\left\{ \begin{array}{l} d_i \cdot p_{ij} = d_j \cdot p_{ji} \\ p_{ij} = f(i, j), \quad p_{ji} = f(j, i) \end{array} \right. \Rightarrow p_{ij} = \sqrt{\frac{d_j}{d_i}}$$

related to def
of graph Laplacians!

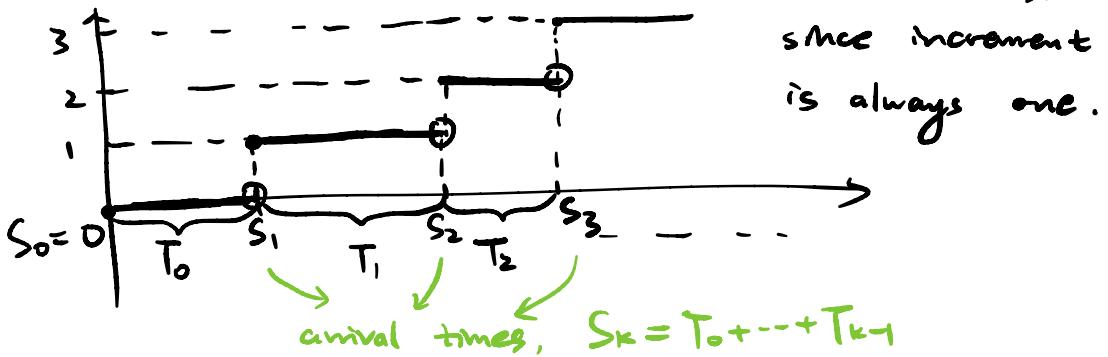
For a review of cts-time Markov Chain, refer to the notes for 213A in fall 2023 on my website.



Check: $Y_t \stackrel{?}{=} X_{N_t}$ for Poisson process $\{N_t\}$ is a Markov Chain.
(intensity λ fixed)
 $\{N_t\} \perp \{X_n\}$



All interarrival times T_0, T_1, \dots i.i.d. $\sim E(\lambda)$.



$\{T_i\}$ also called "holding times" in CTMC

Then $Z_n \triangleq$ the value of X after n -th transition $= X_{S_n}$

so that $Y_t = \sum_{n=0}^{\infty} Z_n \cdot I_{[S_n, S_{n+1})}(t)$,

Lemma: (for Markov)

$$P(Z_{n+1}=j, T_n > u | Z_0, \dots, Z_n, S_1, \dots, S_n)$$

$$= P(Z_{n+1}=j, T_n > u | Z_0, \dots, Z_n, T_0, \dots, T_{n-1})$$

$$= P\left(Z_{n+1}=j, T_n > u | Z_n, T_0, \dots, T_{n-1} \quad \middle| \quad Z_0, \dots, Z_{n-1}\right)$$

Markov $Z_{n+1}|_{Z_n} \perp\!\!\!\perp (Z_0, \dots, Z_{n-1}), \quad \{T_n\} \perp\!\!\!\perp \{Z_0, \dots, Z_{n-1}\}$

$$= P(Z_{n+1}=j, T_n > u | Z_n, T_0, \dots, T_{n-1})$$

\uparrow indep \uparrow indep \uparrow indep

$$= P(Z_{n+1}=j, T_n > u | Z_n) = P_{Z_n, j} \cdot e^{-\lambda u}$$

a func only of Z_n

More generally, holding times only need to be independent, not i.i.d.

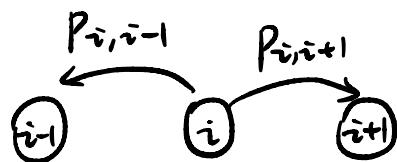


$q(\cdot)$ as holding rate for each state

$$E_1, E_2, \dots \stackrel{\text{i.i.d.}}{\sim} \mathcal{E}(1), T_i = \frac{E_i}{q(z_i)}$$

The lemma still holds, proof modified to include E_i .

e.g: $\{X_n\}$



given $Y_t=i$, B_i and D_i are time until next birth/death.

$$B_i \sim \mathcal{E}(\lambda_i), D_i \sim \mathcal{E}(\mu_i)$$



birth/death rates

$$\text{holding rate } q_i = \lambda_i + \mu_i, \quad \lambda_i = P_{i,i+1} q_i, \quad \mu_i = P_{i,i-1} q_i$$

$$\text{since } T_i = \min\{B_i, D_i\} \sim \mathcal{E}(\lambda_i + \mu_i)$$

e.g.: (cts-time branching with immigration)

Each particle indep. waits $\mathcal{E}(q)$ and either splits into 2 w.p. p or vanishes w.p. $1-p$. New particles immigrate into the system with Poisson arrival, intensity λ . $Y_t \stackrel{\Delta}{=} \#$ of particles in system at time t .

Df:

Given $Y_t = i$, $L_1, \dots, L_i \stackrel{iid}{\sim} \mathcal{E}(q)$ are decision times and $I \sim \mathcal{E}(\lambda)$ is next arrival of immigrants.

Holding time $T_i = \min\{L_1, \dots, L_i, I\} \sim \mathcal{E}(iq + \lambda)$

$$\text{so } \underline{q_i = iq + \lambda}$$

Birth time $B_i = \min\{ \quad , I \} \sim \mathcal{E}(ipq + \lambda)$

$$\text{so } \underline{\lambda_i = ipq + \lambda}, \underline{\mu_i = q_i - \lambda_i = i(1-p)q}$$

thinning
w.p. p

e.g.: (M/M/1 queue)

$$\text{CTMC, BDC, } \begin{cases} \lambda_i = \lambda \\ \mu_i = \mu \\ q_i = \lambda + \mu \end{cases}$$

e.g.: (M/M/ ∞ queue)

$$\text{CTMC, BDC, } \begin{cases} \lambda_i = \lambda \\ \mu_i = i\mu \\ q_i = \lambda + i\mu \end{cases}$$

Regularity of CTMC: $\forall i \in S, \lim_{n \rightarrow \infty} P_i(S_n = \infty) = 1$



$$\forall i \in S, \lim_{n \rightarrow \infty} \left(\sum_n \frac{1}{q_i(x_n)} = \infty \right) = 1$$

e.g.: M/M/1 queue and Poisson process regular.