6. Continuous-Time Markov Chain (cont'd)

6.1. Definitions and Structures (cont'd)

Definition 6.1.1. Continuous-time Stochastic Process (cont'd)

Example 6.1.1.1.

We have seen that the Poisson process satisfy the continuous-time Markov property \Rightarrow it is a CTMC

$$\lambda_i = \lambda \qquad i \in S = \{0,1,2,\cdots\}$$

(interarrival times do not depend on current state. ⇒ time spent in different states are i.i.d.)

$$q_{i,i+i} = 1$$
 $q_{ij} = 0$ otherwise $(j \neq i+1)$

6.2. Generator Matrix

Similar to the discrete-time case, we have the transition probability at time t.

$$egin{aligned} P_{ij}(t) &= \mathbb{P}(X(t) = j | X(0) = i) \ &= \mathbb{P}(X(t+s) = j | X(s) = i) \end{aligned} \quad ext{ assume the MC is time-homogeneous}$$

and matrix

$$P(t) = \{P_{ij}(t)\}_{i,j \in S}$$
 $P(t) = egin{pmatrix} p_{00}(t) & p_{01}(t) & \cdots \ p_{10}(t) & p_{11}(t) & \cdots \ dots & dots \end{pmatrix}$

The C-K equation still holds

$$p(t+s) = P(t) \cdot P(s)$$

Proof: $\forall i,j \in S$

$$egin{aligned} P_{ij}(t+s) &= \mathbb{P}(X(t+s) = j | X(0) = i) \ &= \sum_{k \in S} \mathbb{P}(X(t+s) = j | X(t) = k, X(0) = i) \cdot \mathbb{P}(X(t) = k | X(0) = i) \ &= \sum_{k \in S} \mathbb{P}(X(s) + j | X(o) = k) \cdot \mathbb{P}(X(t) = k | X(0) = i) \ &= \sum_{k \in S} P_{kj}(s) \cdot P_{ik}(t) \ &= (P(t) \cdot P(s))_{ij} \end{aligned}$$

Note that we have

$$P(0)=I$$
 $(P_{ii}(0)=\mathbb{P}(X(0)=i|X(0)=i)=1, P_{ij}(0)=0 ext{ for } j{
eq}i)$

and

$$\lim_{t o 0^+} P(t) = I$$

Actually, we have the following stronger result:

$$R := \lim_{h o 0^+} rac{P(h) - P(0)}{h} = \lim_{h o 0^+} rac{P(h) - I}{h}$$

exists, and is called the (infinitesimal) generator matrix of $\{X(t)\}_{t\geq 0}$

Entry-wise:

$$R_{ij} = \lim_{h o 0^+}rac{P_{ij}(h)-P_{ij}(0)}{h} = egin{cases} \lim_{h o 0^+}rac{P_{ii}(n)-1}{h} \leq 0 & j=i \ \lim_{h o 0^+}rac{P_{ii}(n)}{h} \geq 0 & j
eq i \end{cases}$$

Relation between R and $\{\lambda_i\}_{i\in S}$ and $Q=\{q_{ij}\}_{i,j\in S}$

$$R_{ii} = -\lambda_i \quad , \quad R_{ij} = \lambda_i q_{ij} \qquad j {
eq} i$$

Reason:

$$R_{ii} = \lim_{h
ightarrow 0^+} rac{P_{ii}(h)-1}{h} = \lim_{h
ightarrow 0^+} rac{\mathbb{P}(T_i>h)-1}{h}$$

Where T_i is the random time the process stays in i. The equality holds because when h is very small, the probability of having two or more jumps in time h is negligible.

 $\mathbb{P}(X(h) = i | X(0) = i) = \mathbb{P}(T_i > h) + o(h) \leftarrow ext{having at least 2 jumps and back to } i$

$$*a(h) = o(h) ext{ if } \lim_{h o 0} rac{a(h)}{h} = 0$$