

## Introduction

Schedule:

- (I) Basic properties of continuous-time Markov Chains
- (II) Qualitative properties of continuous time Markov Chains
- (III) Queueing theory
- (IV) Renewal theory
- (V) Spatial Poisson processes

## 1 Some basic aspects of continuous-time Markov Chains

**Definition.** A sequence of random variables is called a *stochastic process* or *process*. The process  $X = (X_n)_{n \geq 1}$  is called a discrete-time Markov Chain with state space  $I$  if for all  $x_0, x_1, \dots, x_n \in I$

$$\mathbb{P}(X_n = x_n | X_{n-1} = x_{n-1}, \dots, X_0 = x_0) = \mathbb{P}(X_n = x_n | X_{n-1} = x_{n-1}).$$

If  $\mathbb{P}(X_{n+1} = y | X_n = x)$  is independent of  $n$ , the chain is called *time-homogeneous*. We then write  $P = (P_{x,y})_{x,y \in I}$  for the *transition matrix* where  $P_{x,y} = \mathbb{P}(X_1 = y | X_0 = x)$ . The data associated to every time-homogeneous Markov Chain is the transition matrix  $P$  and the initial distribution  $\mu$ , i.e  $\mathbb{P}(X_0 = x_0) = \mu(x_0)$ .

From now on:

- $I$  denotes a countable (or finite) state space.
- $(\Omega, \mathcal{F}, \mathbb{P})$  is the probability space on which all the relevant random variables are defined.

**Definition.**  $X = (X(t) : t \geq 0)$  is a (right-continuous) continuous-time random process with values in  $I$  if

- (a) for all  $t \geq 0$ ,  $X(t) = X_t$  is a random variable such that  $X(t) : \Omega \rightarrow I$ ;
- (b) for all  $\omega \in \Omega$ ,  $t \mapsto X_t(\omega)$  is right-continuous (right-continuous sample path).  
In our case this means for all  $\omega \in \Omega$ , for all  $t \geq 0$ , there exists  $\varepsilon > 0$  (depending on  $\omega, t$ ) such that

$$X_t(\omega) = X_s(\omega) \quad \forall s \in [t, t + \varepsilon].$$

**Fact.** A right-continuous random process is defined by its finite-dimensional distributions

$$\mathbb{P}(X_{t_0}=i, X_{t_1}=i_1, \dots, X_{t_n}=i_{t_n}), \quad n \geq 0, \quad t_k \geq 0, \quad i_k \in I.$$

For every  $\omega \in \Omega$ , the path  $t \mapsto X_t(\omega)$  of a right-continuous process stays constant for a while. So there are 3 possibilities:

- (i) The path makes infinitely many jumps overall but only finitely many in a given interval  $[0, t]$ .
- (ii) The path makes finitely many jumps & then gets absorbed in some state.
- (iii) The path makes infinitely jumps in a finite time interval. After the ‘explosion time’  $\zeta$ , the process starts up again.

Write  $J_0 = 0, J_1, J_2, \dots$  for the jump times and  $S_1, S_2, \dots$  for the holding times, defined by

$$J_0 = 0, \quad J_{n+1} = \inf\{t \geq J_n : X_t \neq X_{J_n}\},$$

$$S_n = \begin{cases} J_n - J_{n-1} & J_{n-1} < \infty \\ \infty & \text{otherwise} \end{cases}.$$

By right-continuity,  $S_n > 0$  for all  $n$ . If  $J_{n+1} = \infty$  for some  $n$ , we define  $X_\infty = X_{J_n}$  as the final value, otherwise  $X_\infty$  is not defined. The explosion time  $\zeta$  is defined by

$$\zeta = \sup(J_n) = \sum_{n=1}^{\infty} S_n.$$

We are not going to consider what happens to a chain after explosion. We thus set  $X_t = \infty$  for all  $t \geq \zeta$  (adjoining a new state ‘ $\infty$ ’). We call such a chain *minimal*.

**Definition.** We define the *jump chain*  $Y_n$  of  $(X_t)_{t \geq 0}$  by setting  $Y_n = X_{J_n}$  for all  $n$ .

**Definition.** A right-continuous random process  $X = (X_t)_{t \geq 0}$  has the Markov property (and is called a continuous-time markov chain) if for all  $i_1, i_2, \dots, i_n \in I$  and  $0 \leq t_1 < t_2 < \dots < t_n$ ,

$$\mathbb{P}(X_{t_n} = i_n | X_{t_{n-1}} = i_{n-1}, \dots, X_{t_0} = i_0) = \mathbb{P}(X_{t_n} = i_n | X_{t_0} = i_0).$$

**Remark.** For all  $h > 0$ ,  $Y_n = X(hn)$  defines a discrete-time Markov Chain.

**Definition.** The transition probabilities are  $P_{ij}(s, t) = \mathbb{P}(X_t = j | X_s = i)$ ,  $s \leq t$ ,  $i, j \in I$ . It is called *time-homogeneous* if it depends on  $t - s$  only, i.e

$$P_{ij}(s, t) = P_{i,j}(0, t - s).$$

In this case we just write  $P_{ij}(t - s)$ . As in the case of discrete time, a (time-homogeneous) Markov process is characterised by

1. Its initial distribution  $\lambda_i = \mathbb{P}(X_0 = i)$ ,  $i \in I$ ;
2. Its *family of transition matrices*  $(P(t))_{t \geq 0} = (P_{ij}(t))_{t \geq 0}$ .

The family  $(P(t))_{t \geq 0}$  is called the *transition subgroup* of the MC.