# **Continuous-Time Markov Chains (2)**

### • Alternate definition and Properties of a Poisson Process

- A continuous-time stochastic process  $\{N(t), t \ge 0\}$  counting the number of events (e.g. arrivals) by time t is said to be a Poisson process with rate  $\lambda > 0$ , if
  - (i) N(0) = 0.
  - (ii) N(t) has independent increments. The number of events that occurs in disjoint time intervals are independent.
  - (iii) The number of events that occur in a time interval of length t is Poisson distributed with mean  $\lambda t$ .
- ➤ Obviously, since N(t) counts the number of events, then N(t) takes on nonnegative integer values and  $N(t) \ge N(s)$  for t > s.
- $\triangleright$  Because a Poisson process is also a pure birth process with birth rates all equal to  $\lambda$ , the inter-event time is exponentially distributed with mean  $1/\lambda$ .
- Suppose each event in a Poisson process, N(t), with rate  $\lambda$  can be classified into type I w.p. p and type II w.p. 1-p.
- Then, the number of type I and type II,  $N_1(t)$  and  $N_2(t)$  are independent Poisson processes with rates  $\lambda p$  and  $\lambda(1-p)$ .
- ➤ This is the *decomposition property* of the Poisson process.

#### • Example 5

- ➤ Suppose cars arrive to a gas station according to a Poisson process with rate 5 per hour.
- ➤ What is the probability that 3 cars arrive in an hour?
- Number of cars in an hour, N(1), is Poisson distributed with mean 5. Then,

$$P{N(1) = 3} = e^{-5}(5)^3/3! = 0.14$$
.

- ➤ What is the probability that 2 cars arrive in 15 minutes?
- Number of cars in an 15 minutes, N(1/4), is Poisson distributed with mean 5/4 =. Then,

$$P{N(1/4) = 2} = e^{-5/4}(5/4)^2/2! = 0.224.$$

- ➤ What is the expected time before that the third car in an hour arrives?
- ➤ Inter-arrival times are exponential with mean 1/5 hours. Then, with the memoryless property, the expected time till third is 3/5 hours = 36 minutes.
- ➤ What is the probability that the station, starting empty, has no cars for 30 minutes after opening?
- Let A be the inter-arrival time. A is exponential with mean 1/5 hours. The desired probability is

$$P{A > 1/2} = e^{-5/2} = 0.082.$$

## • Example 6

- $\triangleright$  Customers arrive to a system according to a Poisson process with rate  $\lambda$  and if each customer is a man w.p. 0.5 and a woman w.p. 0.5.
- > Characterize the arrival process of men into the system.
- $\triangleright$  It's a Poisson process with rate  $0.5\lambda$ .

#### • Example 7

- $\triangleright$  Cars arrive to an intersection according to a Poisson process with rate  $\lambda$ . A policeman blocks one way and directs the cars to the other way. On average, the policeman directs half of the cars to street A and the other half to street B.
- ➤ Characterize the arrival process of cars into street A. Is it Poisson?
- ➤ It's not a Poisson process.

### • Limiting Probabilities

➤ Similar to the discrete cases, we define limiting probabilities for a CTMC as

$$P_{j} = \lim_{t \to \infty} P_{ij}(t).$$

 $\triangleright$  Similar to the discrete case also, these probabilities can be interpreted as the long-run fraction of time spent in state j.

- The limiting probabilities exist under the following conditions: (i) all states communicate; and (ii) all states are *positive recurrent* meaning that the expected time to return to a state upon leaving it is finite.
- Assuming that the limiting probabilities exist, they can be determined by Kolmogorov's equations (Theorem 1).
- $\triangleright$  Letting  $t \to \infty$  in Theorem 1, implies that

$$\lim_{t\to\infty} \frac{\partial P_{ij}(t)}{\partial t} = \sum_{k\neq j} q_{kj} \lim_{t\to\infty} P_{ik}(t) - v_j \lim_{t\to\infty} P_{ij}(t) \Rightarrow 0 = \sum_{k\neq j} q_{kj} P_k - v_j P_j$$

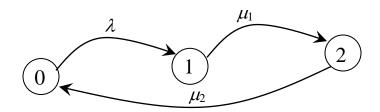
> Therefore,

$$v_j P_j = \sum_{k \neq j} q_{kj} P_k$$

- This equation has an interesting and useful interpretation. The left hand side,  $v_j P_j$ , is the flow out of state j, and the right hand side,  $\sum_{k \neq j} q_{kj} P_k$ , is the flow into state j.
- $\triangleright$  This is a *flow balance equation* (flow out = flow in).

#### Example 8

➤ For the shoeshine shop of Example 1, what is the fraction of time that the shop is busy?



- $\triangleright$  The desired probability is  $P_1 + P_2$  or  $1 P_0$ .
- ➤ The limiting probabilities are given by the following flow balance equations.

State 0. 
$$\lambda P_0 = \mu_2 P_2$$
.

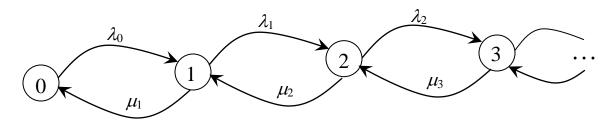
State 1. 
$$\mu_1 P_1 = \lambda P_0$$
.

State 2. 
$$\mu_2 P_2 = \mu_1 P_1$$
.

- Therefore,  $P_1 = (\lambda/\mu_1)P_0$ ,  $P_2 = (\lambda/\mu_2)P_0$ . Noting that  $P_0 + P_1 + P_2 = 1$  implies that  $P_0 = 1/[1 + (\lambda/\mu_1) + (\lambda/\mu_2)]$ .
- > The desired probability is

$$1 - P_0 = \left[ (\lambda/\mu_1) + (\lambda/\mu_2) \right] / \left[ 1 + (\lambda/\mu_1) + (\lambda/\mu_2) \right].$$

### • Limiting Probabilities for a Birth-Death Process



> The flow balance equations are

State 0. 
$$\lambda_0 P_0 = \mu_1 P_1$$
.

State 1. 
$$\mu_1 P_1 + \lambda_1 P_1 = \lambda_0 P_0 + \mu_2 P_2$$
.

State 2. 
$$\mu_2 P_2 + \lambda_2 P_2 = \lambda_1 P_1 + \mu_3 P_3$$
.

State 
$$n > 2$$
.  $\mu_n P_n + \lambda_n P_n = \lambda_{n-1} P_{n-1} + \mu_{n+1} P_{n+1}$ .

- $\triangleright$  Replacing the first equation in the second gives  $\lambda_1 P_1 = \mu_2 P_2$ .
- $\triangleright$  Replacing this in the third equation gives  $\lambda_2 P_2 = \mu_3 P_3$ .

> Continuing in this manner we find that

$$\lambda_n P_n = \mu_{n+1} P_{n+1}, n \geq 0$$

> Then,

$$P_1 = \frac{\lambda_0}{\mu_1} P_0$$
,  $P_2 = \frac{\lambda_1}{\mu_2} P_1 = \frac{\lambda_1 \lambda_0}{\mu_2 \mu_1} P_0$ ,  $P_3 = \frac{\lambda_2}{\mu_3} P_2 = \frac{\lambda_2 \lambda_1 \lambda_0}{\mu_3 \mu_2 \mu_1} P_0$ .

➤ In general,

$$P_n = \frac{\lambda_{n-1}\lambda_{n-2}\dots\lambda_1\lambda_0}{\mu_n\mu_{n-1}\dots\mu_2\mu_1} P_0.$$

ightharpoonup Since  $\sum_{n=0}^{\infty} P_n = 1$ , it follows that

$$P_0 = \left(1 + \sum_{n=1}^{\infty} \frac{\lambda_{n-1} \lambda_{n-2} \dots \lambda_1 \lambda_0}{\mu_n \mu_{n-1} \dots \mu_2 \mu_1}\right)^{-1},$$

$$P_{n} = \frac{\lambda_{n-1}\lambda_{n-2}\dots\lambda_{1}\lambda_{0}}{\mu_{n}\mu_{n-1}\dots\mu_{2}\mu_{1}} \left(1 + \sum_{n=1}^{\infty} \frac{\lambda_{n-1}\lambda_{n-2}\dots\lambda_{1}\lambda_{0}}{\mu_{n}\mu_{n-1}\dots\mu_{2}\mu_{1}}\right)^{-1}, n = 1, 2, \dots$$

 $\triangleright$  A necessary condition for the existence of  $P_n$ s is

$$\sum_{n=1}^{\infty} \frac{\lambda_{n-1}\lambda_{n-2}\dots\lambda_1\lambda_0}{\mu_n\mu_{n-1}\dots\mu_2\mu_1} < \infty.$$