

Continuous-time Markov chains

Stat 568
4-20-17

①

Defn: $\{X(t), t \geq 0\}$ is a continuous-time Markov chain if $\forall s, t \geq 0$ &

integers $i, j, k, u \geq 0, 0 \leq u \leq s,$

$$\begin{aligned} P[X(t+s) = j \mid X(s) = i, X(u) = k(u)] \\ = P[X(t+s) = j \mid X(s) = i] \end{aligned}$$

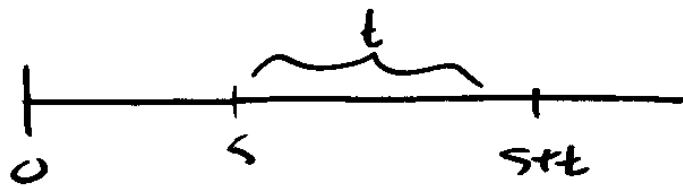
In addition, if

②

$P[X(t+s) = j \mid X(s) = i]$ is independent of s , then the Markov chain has stationary transition probabilities (homogeneous)

Let T_i = time the process stays in state i before making a transition.

$$P[T_i > s+t | T_i > s] = P[T_i > t] \quad (3)$$



This is called the "memory-less" property

Suppose that the random variable T has the memoryless property.

$$\text{Let } \bar{F}(t) = P(T > t)$$

The memoryless property says

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$$\frac{P[T > s+t \mid \cancel{T > s}]}{P[T > s]} = P[T > t]$$

$$P[T > s+t] = P[T > s] P[T > t]$$

$$\bar{F}(s+t) = \bar{F}(s) \bar{F}(t)$$

Consider $g(s+t) = g(s)g(t)$ for continuous g

Let $x = \frac{m}{n}$, where m, n are integers

$$\begin{aligned}
 g\left(\frac{m}{n}\right) &= g\left(\underbrace{\frac{1}{n} + \dots + \frac{1}{n}}_{n \text{ times}}\right) \\
 &= g\left(\frac{1}{n}\right) \cdots g\left(\frac{1}{n}\right) = \left(g\left(\frac{1}{n}\right)\right)^m
 \end{aligned}
 \tag{5}$$

$$\begin{aligned}
 \text{So } \left(g\left(\frac{m}{n}\right)\right)^n &= \left(g\left(\frac{1}{n}\right)\right)^{mn} \\
 &= \left(\left(g\left(\frac{1}{n}\right)\right)^n\right)^m \\
 &= \left(g\left(\underbrace{\frac{1}{n} + \dots + \frac{1}{n}}_{n \text{ times}}\right)\right)^m \\
 &= (g(1))^m
 \end{aligned}$$

$$\text{And } g\left(\frac{m}{n}\right) = (g(1))^{\frac{m}{n}}
 \tag{6}$$

$$g(x) = (g(1))^x \text{ for any real } x \geq 0$$

because x can be sandwiched between 2 converging sequences of rational numbers.

$$\text{Let } \lambda = -\ln g(1) \quad \therefore g(1) = e^{-\lambda}$$

$$g(x) = e^{-\lambda x}$$

$$\therefore \bar{F}(t) = e^{-\lambda t} \quad (7)$$

$$F(t) = 1 - e^{-\lambda t}$$

$$f(t) = \lambda e^{-\lambda t}$$

\therefore In a continuous-time Markov chain, the time spent in state i before transitioning is exponentially distributed.

Defn. let $\{X(t), t \geq 0\}$ be a continuous time Markov chain (8)

When $X(t) = n$,

i) new arrivals enter at exponential rate λ_n

ii) departures occur at exponential rate μ_n

Then $X(t)$ is called a birth and death process

And $\{\lambda_n\}_{n=0}^{\infty}$ and $\{\mu_n\}_{n=0}^{\infty}$ are the arrival & departure rates

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Let ν_i be the exponential parameter of T_i

From state 0, the process can only go to state 1.

$$\text{So } P_{01} = 1 \quad \nu_0 = \text{parameter of } T_0 \\ = \lambda_0$$

For $i \geq 1$,

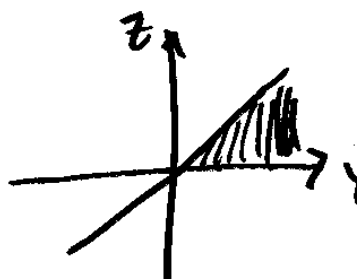
the process can only go from state i to
state $i-1$ or $i+1$

To go from i to $i-1$, a death must occur
before a birth.

(10)

Let $Y \sim \exp(\lambda)$ and $Z \sim \exp(\mu)$, indep.
(birth) (death)

$$P(Z < Y) = \int_0^\infty \int_0^y \lambda e^{-\lambda y} \mu e^{-\mu z} dz dy$$



$$= \int_0^\infty \lambda e^{-\lambda y} \left(-e^{-\mu z} \right) \Big|_{z=0}^y dy$$

$$= \int_0^\infty \lambda e^{-\lambda y} \left(-e^{-\mu y} - (-1) \right) dy$$

$$= \int_0^\infty \lambda \left(e^{-\lambda y} - e^{-(\lambda+\mu)y} \right) dy$$

$$= \lambda \left[\frac{e^{-\lambda y}}{-\lambda} - \frac{e^{-(\lambda+\mu)y}}{-(\lambda+\mu)} \right]_{y=0}^{\infty} \quad (11)$$

$$= \lambda \left[0 - \left(-\frac{1}{\lambda} + \frac{1}{\lambda+\mu} \right) \right]$$

$$= 1 - \frac{\lambda}{\lambda+\mu} = \frac{\mu}{\lambda+\mu} = P_{i,i-1}$$

$$\text{So } P_{i,i+1} = \frac{\lambda}{\lambda+\mu}$$

$$\text{So far: } P_{0,1} = 1 \quad (12)$$

$$P_{i,i-1} = \frac{\mu_i}{\lambda_i + \mu_i}$$

$$P_{i,i+1} = \frac{\lambda_i}{\lambda_i + \mu_i}$$

T_i is time in state i , so T_i is the minimum
of (time until birth, time until death)

Again $Y \sim \exp(\lambda)$, $Z \sim \exp(\mu)$ indep.

(13)

Let $T = \min(Y, Z)$

$$P(T > t) = P(\min(Y, Z) > t)$$

$$= P(Y > t \text{ and } Z > t)$$

$$= P(Y > t) P(Z > t)$$

$$= e^{-\lambda t} e^{-\mu t}$$

$$= e^{-(\lambda + \mu)t}$$

So T is $\exp(\lambda + \mu)$ $\therefore \nu_i = \lambda_i + \mu_i$

Example Linear growth with immigration

(14)

Assume $\lambda_n = n\lambda + \theta$

and $\mu_n = n\mu$

Find $E[X(t)]$

Let $M(t) = E[X(t)]$

So $M(t+h) = E[X(t+h)]$

$$= E[E[X(t+h) | X(t)]]$$

(15)

Given $X(t)$,

Probability:

$$X(t+h) = \begin{cases} X(t)+1 & (X(t)\lambda + \theta)h + o(h) \\ X(t)-1 & (X(t)\mu)h + o(h) \\ X(t) & 1 - \text{other 2} \end{cases}$$

$$E[X(t+h) | X(t)] =$$

$$\begin{aligned} & (X(t)+1) [(X(t)\lambda + \theta)h + o(h)] \\ & + (X(t)-1) [X(t)\mu h + o(h)] \\ & + X(t) [1 - (X(t)\lambda + \theta)h - X(t)\mu h + o(h)] \end{aligned}$$

$$= X(t) + [(X(t)\lambda + \theta)h - X(t)\mu h + o(h)] \quad (16)$$

$$M(t+h) = E(\cdot)$$

$$= M(t) + M(t)(\lambda - \mu)h + \theta h + o(h)$$

$$\frac{M(t+h) - M(t)}{h} = M(t)(\lambda - \mu) + \theta + \frac{o(h)}{h}$$

Take $\lim_{h \rightarrow 0}$

$$M'(t) = (\lambda - \mu)M(t) + \theta$$

$$\text{Let } g(t) = (\lambda - \mu)M(t) + \theta$$

$$g'(t) = (\lambda - \mu)M'(t)$$

$$\text{So } M'(t) = \frac{g'(t)}{\lambda - \mu}$$

$$\frac{g'(t)}{\lambda - \mu} = g(t)$$

$$\frac{g'(t)}{g(t)} = \lambda - \mu$$

$$\ln g(t) = (\lambda - \mu)t + c$$

$$g(t) = ke^{(\lambda - \mu)t}$$

$$ke^{(\lambda - \mu)t} = (\lambda - \mu)M(t) + \theta$$

$$M(0) = E[X(0)] = X(0) = x_0$$

$$k = (\lambda - \mu)x_0 + \theta$$

$$\text{Finally } M(t) = \frac{ke^{(\lambda - \mu)t} - \theta}{\lambda - \mu} = \dots$$

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HW#3

6. Consider a birth and death process with birth rates $\lambda_i = (i + 1)\lambda, i \geq 0$, and death rates $\mu_i = i\mu, i \geq 0$.
- (a) Determine the expected time to go from state 0 to state 4.
 - (b) Determine the expected time to go from state 2 to state 5.
 - (c) Determine the variances in parts (a) and (b).
9. The birth and death process with parameters $\lambda_n = 0$ and $\mu_n = \mu, n > 0$ is called a pure death process. Find $P_{ij}(t)$.
12. Each individual in a biological population is assumed to give birth at an exponential rate λ , and to die at an exponential rate μ . In addition, there is an exponential rate of increase θ due to immigration. However, immigration is not allowed when the population size is N or larger.
- (a) Set this up as a birth and death model.
 - (b) If $N = 3, 1 = \theta = \lambda, \mu = 2$, determine the proportion of time that immigration is restricted.
17. Each time a machine is repaired it remains up for an exponentially distributed time with rate λ . It then fails, and its failure is either of two types. If it is a type 1 failure, then the time to repair the machine is exponential with rate μ_1 ; if it is a type 2 failure, then the repair time is exponential with rate μ_2 . Each failure is, independently of the time it took the machine to fail, a type 1 failure with probability p and a type 2 failure with probability $1 - p$. What proportion of time is the machine down due to a type 1 failure? What proportion of time is it down due to a type 2 failure? What proportion of time is it up?