Lecture 16

对于给定 states i, j, 希望知道 transition probabilities Pij(t), サセシロ

- 31 Chapman Kolmogorov equation
- 1. Theorem: Chapman-Kolmogorov equation (C-K方程)

Chapman - Kolmogorov equation \$:

$$Pij(t+s) = \sum_{k \in Pik} Pik(t) Pkj(s)$$
, $t > 0$, $i,j \in E$

选: CK方程表示:

如 明:

$$Pij(t+s) = P\{X(t+s) = j \mid X(0) = i\}$$

$$= \sum_{k \in L} P\{X(t+s) = j, X(t) = k \mid X(0) = i\}$$

$$= \sum_{k \in L} P\{X(t+s) = j \mid X(t) = k, X(0) = i\} \cdot P\{X(t) = k \mid X(0) = i\}$$

$$= \sum_{k \in L} P_{ik}(t) P_{kj}(s)$$

2、CK方程的矩阵表示

根据C-K方程,有

进一步可推出:

$$P(2t) = P(t+t) = P(t)^2$$

证明:

$$P_{ij}(t+s) = \sum_{k \in E} P_{ik}(t) P_{kj}(s)$$

$$= [P_{i1}(t), P_{in}(t), \cdots, P_{i|E|}(t)] \cdot [P_{ij}(s), P_{nj}(s), \cdots, P_{i|E|}(s)]^{T}$$

$$= i - th \text{ row of } P(t) \cdot j - th \text{ column of } P(s)$$

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

eg. ▶ Suppose

$$P(0.1) = \begin{pmatrix} 0.7486327 & 0.1607327 & 0.0906346 \\ 0.0783127 & 0.8310527 & 0.0906346 \\ 0.0041073 & 0.0865273 & 0.9093654 \end{pmatrix}$$

Compute

$$\mathbb{P}\{X(.4) = 3, X(.2) = 1, X(.1) = 3 | X(0) = 2\}$$

$$= (P(0.1))_{1,3}^{2} P_{3,1}(0.1) P_{2,3}(.1)$$

$$= (0.164840)(0.0041073)(0.0906346).$$

§2 Transition probability matrix 的末解

1. Transition probability matrix 的末瓣 (Generator matrix 的 motivation)

全G为 the generator matrix, 因

$$P(s) = \sum_{k=0}^{\infty} \frac{s^k G^k}{k!}$$
, $\sharp P G^{\circ} = I$

证明

证明: (#)式

①当沙j时,有

$$P_{ij}(0+) = \lim_{t \to 0} \frac{P_{ij}(t) - P_{ij}(0)}{t}$$

$$= \lim_{t \to 0} \frac{P(X(t) = j \mid X(0) = i)}{t} \quad (P_{ij}(0) = 0, P_{ij}(t) = P(X(t) = j \mid X(0) = i))$$

$$\approx \lim_{t \to 0} \frac{P(\text{one jump in } [0, t) \text{ from } i \text{ to } j)}{t} \quad (t \to 0 \text{ M}, [0, t) \text{ M$\frac{1}{3}$} - \lambda_j \text{ jump})$$

$$= \lim_{t \to 0} \frac{P(\text{one jump in } [0, t)) \cdot P(i \to j \mid \text{one jump})}{t}$$

$$= \lim_{t \to 0} \frac{P(\text{Exp}(\lambda_i) < t) \cdot J_{ij}}{t}$$

$$= \lim_{t \to 0} \frac{(1 - e^{-\lambda(i)t}) \cdot J_{ij}}{t}$$

$$= \lim_{t \to 0} \frac{\lambda(i) e^{-\lambda(i)t} \cdot J_{ij}}{1} \quad (L' \text{ H$\^{opital}'s rule})$$

②当证j时,往意到

$$\frac{5}{5}P_{3}^{\prime}(0) = \lim_{t \to 0} \frac{5}{5}P_{3}(t) - \frac{5}{5}P_{3}(0) = \lim_{t \to 0} \frac{0}{t} = 0$$

注: ① Forward equation: Pit1=PIt1G,其与 backward equation 筋解相同.

D Backword equation 水流成主 Forward equation 在 state space 为 infinity 时可能不成立

③ 对于 multiple-states CTMC, 使用 backward equation 求解 P(t) 仍不容易

图1: 机器在根环前的工作时长服从 exponential distribution with mean 方 (Exp(知)) 修理时长服从 exponential distribution with mean 立 (Exp (UI))

若机器在time D 时处于work condition,

求其在 time 七时处于 work condition 的概率

可作出 transition rate diagram

围比 generator matrix G为

$$G = \begin{bmatrix} -\lambda & \lambda \\ \mu & -\mu \end{bmatrix}$$

由 backward equation Pit1=GP(t)可知:

代入 generator matrix,有

$$\begin{cases} P_{00}(t) = G_{01}P_{10}(t) + G_{00}P_{00}(t) = \lambda P_{10}(t) - \lambda P_{00}(t) & (*) \\ P_{10}(t) = G_{10}P_{00}(t) + G_{11}P_{10}(t) = \mu P_{00}(t) - \mu P_{10}(t) & (#) \end{cases}$$

(由于 Part) = 1-Part), Part) = 1-Part), 因此仅需引出两个方程)

· 对 (*) 两侧同乘 M , 对 (#) 两侧同乘 A , 随后相加 , 有:

$$\Rightarrow (\mu P_{00}(t) + \lambda P_{10}(t))' = 0$$

代入七口可得:

因此,

$$P_{10}(t) = \frac{M}{3} (1 - P_{00}(t))$$

代入(*)可知,

$$P_{00}(t) = \lambda P_{00}(t) - \lambda P_{00}(t)$$

$$= \mu (1 - P_{00}(t)) - \lambda P_{00}(t)$$

$$= \mu - (\lambda + \mu) P_{00}(t) \quad (\Delta)$$

由 咖啡知识到知。

$$Poo(t) = C_1 + C_2 e^{-(\lambda + \mu)t}$$

$$-(\lambda+\mu)C_2e^{-(\lambda+\mu)t}=\mu-(\lambda+\mu)[C_1+C_2e^{-(\lambda+\mu)t}]$$

$$\Rightarrow$$
 $C_1 = \frac{\mu}{\lambda + \mu}$

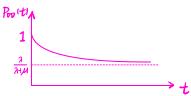
$$1 = \frac{\mu}{\lambda + \mu} + C_{\lambda}$$

$$\Rightarrow C_2 = \frac{\lambda}{\lambda + \mu}$$

因此,

$$P_{00}(t) = \frac{\mu}{\lambda + \mu} + \frac{\lambda}{\lambda + \mu} e^{-(\lambda + \mu)t}$$

注: 加切随时间七的曲线为:



其中 温 为极限概率

Assume $X = \{X(t), t \ge 0\}$ is a CTMC on state space $E = \{1, 2, 3\}$ with generator

$$G = \begin{pmatrix} -3 & 2 & 1 \\ 1 & -2 & 1 \\ 0 & 1 & -1 \end{pmatrix}$$

► Find

$$\mathbb{P}\{X(5.4) = 3, X(2.1) = 1 | X(0) = 2)\} = P_{2,1}(2.7)P_{1,3}(3.3),$$

► Use function expm(·) in *Python* or similar functions in other programming languages,

$$P(2.7) = e^{2.7G} = \begin{pmatrix} 0.12614 & 0.37612 & 0.49774 \\ 0.12612 & 0.37614 & 0.49774 \\ 0.12387 & 0.37387 & 0.50226 \end{pmatrix}, \quad P(3.3) = e^{3.3G}.$$