

- Uniformization.
 Last time, if all $v_i = v$,
 then conditioning on number of events

$$P_{ij}(t) = \sum_{n=0}^{\infty} P_{ij}^{(n)} e^{-vt} \frac{(vt)^n}{n!}$$
 which can be used to compute numeric
 values for $P_{ij}(t)$.
 The result can also be useful when
 $v_i \leq v \quad \forall i$
 by allowing fictitious transitions from
 a state to itself with probability $1 - \frac{v_i}{v}$.
 Leaving a state with rate v_i can be
 thought as leaving transitions at rate
 v but only $\frac{v_i}{v}$ are real and a fraction
 $1 - \frac{v_i}{v}$ just remain in state i .

Thus a continuous time Markov
 chain with rate v_i where $v_i \leq v \quad \forall i$
 has the same distribution as a Markov
 chain with rate v where the jump
 matrix is
$$P_{ij}^* = \begin{cases} 1 - \frac{v_i}{v} & \text{(fictitious transitions)} \\ \frac{v_j}{v} & \text{(real transitions)} \end{cases}$$

 Then
$$P_{ij}(t) = \sum_{n=0}^{\infty} \binom{vt}{n} P_{ij}^{*(n)} e^{-vt} \frac{(vt)^n}{n!}$$

 Example, machine repair
 state $\begin{matrix} 0 & \text{on} \\ 1 & \text{off} \end{matrix}$ breakdown rate v & repair rate μ
 $P_{01} = P_{10} = 1$
 Take $v = \lambda + \mu$
 Then
$$P_{0j} = \begin{cases} 1 - \frac{\mu}{\lambda + \mu}, j=0 \\ \frac{\mu}{\lambda + \mu} P_{01}, j=1 \\ 1 - \frac{\mu}{\lambda + \mu}, j=1 \\ \frac{\mu}{\lambda + \mu} P_{01}, j=0 \end{cases} = \begin{cases} \frac{\lambda}{\lambda + \mu}, j=0 \\ \frac{\mu}{\lambda + \mu}, j=1 \\ \frac{\lambda}{\lambda + \mu}, j=1 \\ \frac{\mu}{\lambda + \mu}, j=0 \end{cases}$$

$$P_{1j} = \begin{cases} 1 - \frac{\lambda}{\lambda + \mu}, j=1 \\ \frac{\lambda}{\lambda + \mu} P_{10}, j=0 \end{cases} = \begin{cases} \frac{\mu}{\lambda + \mu}, j=1 \\ \frac{\lambda}{\lambda + \mu}, j=0 \end{cases}$$

Hence $P_{00}^{\infty} = P_{10}^{\infty} = \frac{\lambda}{\lambda + \mu}$
 $P_{11}^{\infty} = P_{01}^{\infty} = \frac{\mu}{\lambda + \mu}$
 The transition into 0 (or 1) does not
 depend on the present state, so
 $P_{i0}^{\infty} = \frac{\lambda}{\lambda + \mu}$ and $P_{i1}^{\infty} = \frac{\mu}{\lambda + \mu}$ when $n \geq 1$
 Hence
$$P_{00}(t) = \sum_{n=0}^{\infty} P_{00}^{(n)} e^{-(\lambda + \mu)t} \frac{[(\lambda + \mu)t]^n}{n!}$$

$$= e^{-(\lambda + \mu)t} + \frac{\mu}{\lambda + \mu} \cdot e^{-(\lambda + \mu)t} \sum_{n=1}^{\infty} \frac{[(\lambda + \mu)t]^n}{n!}$$

$$= e^{-(\lambda + \mu)t} \cdot \frac{\lambda}{\lambda + \mu} \cdot [1 - e^{-(\lambda + \mu)t}]$$

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

 as before, for P_{00} in p. 372.

Similarly

$$P_{11}(t) = \frac{\mu}{\lambda + \mu} + \frac{\lambda}{\lambda + \mu} [1 - e^{-(\lambda + \mu)t}]$$

$$P_{01}(t) = 1 - P_{00}(t)$$

$$P_{10}(t) = 1 - P_{11}(t)$$

 Let $O(t)$ = time spent in state 0
 in interval $(0, t)$

$$E[O(t)] = \int_0^t P(X(s) = 0) ds$$

$$= \int_0^t P(X(s) = 0) ds$$

$$= \int_0^t \left[\frac{\lambda}{\lambda + \mu} + \frac{\mu}{\lambda + \mu} e^{-(\lambda + \mu)s} \right] ds$$

$$= \frac{\lambda}{\lambda + \mu} t + \frac{\mu}{(\lambda + \mu)^2} [1 - e^{-(\lambda + \mu)t}]$$

6.9. Computing the transition probabilities
 Introduce
$$r_{ij} = \begin{cases} q_{ij} & i \neq j \\ -v_i & i = j \end{cases}$$

 Kolmogorov backward equation

$$P_{ij}'(t) = \sum_{k \neq i} q_{ik} P_{kj}(t) - v_i P_{ij}(t)$$

$$= \sum_k r_{ik} P_{kj}(t)$$

 forward equations

$$P_{ij}'(t) = \sum_{k \neq j} q_{kj} P_{ik}(t) - v_j P_{ij}(t) = \sum_k r_{kj} P_{ik}(t)$$

 Define matrices $Q, P(t), P'(t)$
 having (i, j) elements equal to $r_{ij}, P_{ij}(t), P_{ij}'(t)$
 Backward equation: $P'(t) = Q P(t)$
 Forward $P'(t) = P(t) Q$
 Q : infinitesimal generator matrix

$f'(t) = f(t) \quad f = c e^{-t}$
 Solution of the equations are

$$P(t) = P(0) \cdot e^{Qt} = e^{Qt} \quad \text{since } P(0) = I$$

 where
$$e^{Qt} = \sum_{n=0}^{\infty} \frac{Q^n t^n}{n!}$$

 Direct use of $P(t) = e^{Qt}$
 can be complicated numerically since
 Q contain negative values.
 But there are alternatives.