

$$P(t) = e^{At}$$

$$R = \begin{pmatrix} -\lambda_0 & \lambda_0 & 0 & \dots \\ 0 & -\lambda_1 & \lambda_1 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

$$R_{ij} = \begin{cases} -\lambda_i, & i=j \\ \lambda_{ij}, & i \neq j \end{cases}$$

$$v = \begin{pmatrix} \text{matrix of} \\ \text{eigenvec} \end{pmatrix}$$

$$P_{ij} = v_i P_{ij}$$

$$R = v^{-1} D v$$

$$D = \begin{pmatrix} \lambda_0 & 0 \\ 0 & \lambda_n \end{pmatrix} \Rightarrow e^{Rt} = v^{-1} e^{Dt} v$$

## Recitation 1

Continuous MC, definition, transition probability

Problem 1.

Consider a birth and death process with birth rates  $\lambda_i = (i+1)\lambda$ ,  $i \geq 0$ , and death rates  $\mu_i = i\mu$ ,  $i \geq 0$ .

- Determine the expected time to go from state 0 to state 4.
- Determine the expected time to go from state 2 to state 5.
- Determine the variances in parts (a) and (b).

$$P'(t) = R P(t) > P(t) e^{Rt}$$

Problem 2

In a birth and death process with birth parameter  $\lambda_n = \lambda$ ,  $n=0,1,\dots$ , and death parameters  $\mu_n = n\mu$

for  $n=0,1,\dots$  we have  $P_{0j}(t) = \frac{(\lambda\mu)^j e^{-\lambda t}}{j!}$

Where

$$p = \frac{1}{\mu} (1 - e^{-\mu t})$$

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$$P'_{0j}(t) = \sum_{k=j}^{\infty} \lambda_{k-1} P_{0,k-1}(t) - (\lambda_j - \mu_j) P_{0j}(t) = \lambda_{j-1} P_{0,j-1}(t) + \mu_{j+1} P_{0,j+1}(t) - (\lambda_j - \mu_j) P_{0j}(t)$$

Verify that these transition probabilities satisfy the forward equation with  $i=0$ .

$$R = \begin{pmatrix} -3\mu & \mu & \mu & \mu \\ \mu & -3\mu & \mu & \mu \\ \mu & \mu & -3\mu & \mu \\ \mu & \mu & \mu & -3\mu \end{pmatrix}$$

Problem 3

(Jukes-Cantor Model). In this chain, the states are the four nucleotides

A, C, G, T. Jumps, which correspond to nucleotide substitutions, occur according to rate  $q_{ij} = \mu$  if  $i \neq j$ . Find the transition probability matrix  $P(t)$  using forward differential equation.

Problem 4

The nucleotides A and G are purines while C's and T's are pyrimidines. Kimura's model takes into account that mutations that do not change the type of base (called transitions) happen at a different rate than those that do (called transversions), so the transition matrix P

$$R = \begin{pmatrix} -(\alpha + 2\beta) & \alpha & \beta & \beta \\ \alpha & -(\alpha + 2\beta) & \beta & \beta \\ \beta & \beta & -(\alpha + 2\beta) & \alpha \\ \beta & \beta & \alpha & -(\alpha + 2\beta) \end{pmatrix}$$

Find  $P(t)$  using forward differential equation.

1. birth rate:  $\lambda_i \cdot (i+1)$ ,  $i \geq 0$   
 death rate:  $\mu_i \cdot i$ ,  $i \geq 0$   
 (a)  $E(i_{0.4}) = ?$

$$E(T_i) = \frac{1}{\lambda_i} + \frac{\mu_i}{\lambda_i} E(T_{i-1})$$

$$S_4 = T_0 + T_1 + T_2 + T_3$$

$$E(S_4), W(S_4)$$

$$W(T_0) = \frac{1}{\lambda_0^2}$$

$$W(T_i) = \frac{1}{\lambda_i(\lambda_i + \mu_i)} + \frac{\mu_i}{\lambda_i} W(T_{i-1}) + \frac{\mu_i}{\mu_i + \lambda_i} [E(T_{i-1}) + E(T_i)]$$

$$E(T_0) = \frac{1}{\lambda}$$

$$E(T_1) = \frac{1}{2\lambda} + \frac{\mu}{2\lambda} \left( \frac{1}{\lambda} \right) = \frac{1}{2\lambda} \left( 1 + \frac{\mu}{\lambda} \right)$$

$$E(T_2) = \frac{1}{3\lambda} + \frac{2\mu}{3\lambda} \left[ \frac{1}{2\lambda} \left( 1 + \frac{\mu}{\lambda} \right) \right] = \frac{1}{3\lambda} \left[ 1 + \frac{\mu}{\lambda} + \left( \frac{\mu}{\lambda} \right)^2 \right]$$

$$E(T_3) = \frac{1}{4\lambda} + \frac{3\mu}{4\lambda} \left\{ \frac{1}{3\lambda} \left[ 1 + \frac{\mu}{\lambda} + \left( \frac{\mu}{\lambda} \right)^2 \right] \right\} = \frac{1}{4\lambda} \left[ 1 + \frac{\mu}{\lambda} + \left( \frac{\mu}{\lambda} \right)^2 + \left( \frac{\mu}{\lambda} \right)^3 \right]$$

$$E(i_{0.4}) = \frac{1}{\lambda} + \frac{1}{2\lambda} \left( 1 + \frac{\mu}{\lambda} \right) + \frac{1}{3} \left[ 1 + \frac{\mu}{\lambda} + \left( \frac{\mu}{\lambda} \right)^2 \right] + \frac{1}{4\lambda} \left[ 1 + \frac{\mu}{\lambda} + \left( \frac{\mu}{\lambda} \right)^2 + \left( \frac{\mu}{\lambda} \right)^3 \right]$$

$$W(T_0) = \frac{1}{\lambda_0^2}$$

$$W(T_1) = \frac{1}{\lambda_1(\lambda_1 + \mu_1)} + \frac{\mu_1}{\lambda_1} W(T_0) + \frac{\mu_1}{\mu_1 + \lambda_1} [E(T_0) + E(T_1)]^2$$

$$= \frac{1}{\lambda_1(\lambda_1 + \mu_1)} + \frac{\mu_1}{\lambda_1} \left( \frac{1}{\lambda_0^2} \right) + \frac{\mu_1}{\mu_1 + \lambda_1} \left[ \frac{1}{\lambda} + \frac{1}{2\lambda} \left( 1 + \frac{\mu}{\lambda} \right) \right]^2$$

$$W(T_2) = \frac{1}{\lambda_2(\lambda_2 + \mu_2)} + \frac{\mu_2}{\lambda_2} W(T_1) + \frac{\mu_2}{\mu_2 + \lambda_2} [E(T_1) + E(T_2)]^2$$

$$W(T_3) = \frac{1}{\lambda_3(\lambda_3 + \mu_3)} + \frac{\mu_3}{\lambda_3} W(T_2) + \frac{\mu_3}{\mu_3 + \lambda_3} [E(T_2) + E(T_3)]^2$$

(b)  $E(i_{1.5})$

$$E(T_4) = \frac{1}{5\lambda} +$$