

An adjacency co-evolutionary model

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1 Introduction to related concepts

Definition 1. Given two adjacencies $a=\{a_1,a_2\}$ and $b=\{b_1,b_2\}$, an adjacency pair is defined as a combination of two adjacencies.

For example: $\langle \{a_1,a_2\},\{b_1,b_2\} \rangle$ is an adjacency pair.

Definition 2. In genome G , we define a certain adjacency pair $\langle a, b \rangle$ to have five states.

For example: $\langle 1, 1 \rangle$ indicates that both adjacencies are present in genome G , $\langle 0, 0 \rangle$ indicates that neither adjacency is present in genome G , $\langle 1, 0 \rangle$ and $\langle 0, 1 \rangle$ indicate the cases where a is present and b is absent and a is absent and b is present, respectively. Particularly, if a and b are both present in genome G but located in different chromosomes, we use $\langle 1, 1 \rangle'$.

2 Description of an adjacency Co-evolutionary Model

2.1 Calculation of the adjacency-pairs probability

To obtain each state of each adjacency pair of a particular node, it is sufficient to be given a description of the structure of the evolutionary tree.

2.1.1 Probabilistic Models

Given a transition rate matrix Q describing the birth-death process.

$$Q = \begin{pmatrix} -3a - b & a & a & a & b \\ a & -3a - b & a & a & b \\ a & a & -3a - b & a & b \\ a & a & a & -3a - b & b \\ b & b & b & b & -4b \end{pmatrix} \quad (1)$$

Where a denotes the transition rate between two adjacency-pairs in the same chromosome and b denotes the transition rate between two adjacency-pairs in different chromosomes. Let the matrix $P(t) = P_{ij}(t)$ be a 5-dimensional matrix, where $P_{ij}(t)$ denotes the probability that state i transitions to state j after time t . $P(t + dt)$ is the state transition matrix at time $t + dt$. We can derive from the rate matrix Q that $P(t + dt) = P(t) + QP(t)dt$. Then we can obtain the differential equation $P'(t) = QP(t)$, and by solving this equation we can further obtain $P(t) = P(0) \times e^{Qt}$, for $P(0) = 1$, we have $P(t) = e^{Qt}$. Then we diagonalize the Q matrix to obtain: $Q = VDU$. where V, D, U are:

$$V = \begin{pmatrix} 1 & 1 & -1 & -1 & -1 \\ 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 1 & -4 & 0 & 0 & 0 \end{pmatrix} \quad (2)$$

$$D = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & -5b & 0 & 0 & 0 \\ 0 & 0 & -4a - b & 0 & 0 \\ 0 & 0 & 0 & -4a - b & 0 \\ 0 & 0 & 0 & 0 & -4a - b \end{pmatrix} \quad (3)$$

$$V = \begin{pmatrix} 1/5 & 1/5 & 1/5 & 1/5 & 1/5 \\ 1/20 & 1/20 & 1/20 & 1/20 & -1/5 \\ -1/4 & -1/4 & -1/4 & -1/4 & 0 \\ -1/4 & -1/4 & 3/4 & -1/4 & 0 \\ -1/4 & -1/4 & -1/4 & 3/4 & 0 \end{pmatrix} \quad (4)$$

Next, we perform a Taylor expansion of the formula $P(t) = e^{Qt}$ to obtain the following transition probability matrix: $p_{ij}(t) = \sum_{k=1}^5 V_{ik} U_{kj} \exp(td_k)$. where V_i is the value of the i th row of the V matrix. U_j is the j th column value of the U matrix. d_k is the k th eigenvalue of the D matrix. So the transition probability matrix is as follows:

$$P = \begin{pmatrix} p_1 & p_2 & p_2 & p_2 & p_3 \\ p_2 & p_1 & p_2 & p_2 & p_3 \\ p_2 & p_2 & p_1 & p_2 & p_3 \\ p_2 & p_2 & p_2 & p_1 & p_3 \\ p_3 & p_3 & p_3 & p_3 & p_4 \end{pmatrix} \quad (5)$$

Where:

$$\begin{aligned} p_1 &= \frac{1}{5} + \frac{1}{20}e^{-5bt} - \frac{3}{4}e^{-4at-bt} \\ p_2 &= \frac{1}{5} + \frac{1}{20}e^{-5bt} - \frac{1}{4}e^{-4at-bt} \\ p_3 &= \frac{1}{5} - \frac{1}{5}e^{-5bt} \\ p_4 &= \frac{1}{5} + \frac{1}{5}e^{-5bt} \end{aligned}$$

2.1.2 Parameter estimation

From the value of p_i we can derive: $p_4 - p_3 = e^{-5bt}$ and $p_1 - p_2 = e^{-4at-bt}$. Accordingly, we can make the following parameter estimates: $bt = -\frac{1}{5} \ln(p_4 - p_3)$, $at = \frac{1}{20} \ln(p_4 - p_3) - \frac{1}{4} \ln(p_1 - p_2)$.

2.1.3 Calculation of the adjacency pairs probability

For an evolutionary tree T with all leaf node genomes known, we complete the collection of adjacency-pairs according to the method in Section 3.2 and perform the screening of adjacency-pairs according to the method in Section 3.3. Based on the filtered list of adjacency-pairs, we encoded five states for each adjacency-pair of each genome separately. Suppose a is an internal node in an evolutionary tree T and a site s is one of the sites in the list of adjacent-pairs. D_a is the observed data for all leaves of the evolutionary tree at site s . We calculate the conditional probability of each state on each site s of the internal node a using the following Bayesian formula:

$$p(s_a|D_a) = \frac{P(s_a)P(D_a|s_a)}{P(D_a)} = \frac{P(s_a)P(D_a|s_a)}{\sum_{s_a} P(s_a)P(D_a|s_a)} = \frac{\pi_{s_a} L_a(s_a)}{\sum_{s_a} \pi_{s_a} L_a(s_a)} \quad (6)$$

$P(s_a)$ is the prior probability estimated from π_{s_a} , while π_{s_a} is the frequency of state s_a in the leaf genome. In the equation, $P(D_a)$ is calculated by the total probability of all states of the site s . $P(D_a|s_a)$ is expressed through $L_a(s_a)$, which is the conditional probability of the state s_a of the site s in the descendant of the internal node a . It is known that $L_a(s_a)$ can be computed recursively by taking a post-order traversal [26], and we compute the value of $L_a(s_a)$ according to the following equation:

$$L_a(s_a) = \begin{cases} 1 & \text{if } a \text{ is a leaf node with state } = s_a \text{ at the site} \\ 0 & \text{if } a \text{ is a leaf node with state } \neq s_a \text{ at the site} \\ \sum_{s_l} p_{s_a s_l} L_l(s_l) \cdot \sum_{s_r} p_{s_a s_r} L_r(s_r) & \text{otherwise} \end{cases} \quad (7)$$

where $p_{s_a s_l}$ is the transition probability of state s_a to state s_l after evolution time t . After this step, we will obtain the probability of each state for each adjacency-pair of each internal node.