### Molecular Evolution

Bret Larget

Departments of Botany and of Statistics University of Wisconsin—Madison

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#### Features of Molecular Evolution

- Possible multiple changes on edges
- Transition/transversion bias
- Non-uniform base composition
- Rate variation across sites
- Opendence among sites
- Codon position
- Protein structure

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### A Famous Quote About Models

Essentially, all models are wrong, but some are useful.

George Box

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## **Probability Models**

- A probabilistic framework provides a platform for formal statistical inference
- Examining goodness of fit can lead to model refinement and a better understanding of the actual biological process
- Model refinement is a continuing area of research
- Most common models of molecular evolution treat sites as independent
- These common models just need to describe the substitutions among four bases at a single site over time.

### The Markov Property

- Use the notation X(t) to represent the base at time t.
- Formal statement:

$$P\{X(s+t) = j \mid X(s) = i, X(u) = x(u) \text{ for } u < s\}$$

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

- Informal understanding: given the present, the past is independent of the future
- If the expression does not depend on the time s, the Markov process is called *homogeneous*.

#### Rate Matrix

- Positive off-diagonal rates of transition
- Negative total on the diagonal
- Row sums are zero
- Example

$$Q = \{q_{ij}\} = \begin{pmatrix} -1.1 & 0.3 & 0.6 & 0.2 \\ 0.2 & -1.1 & 0.3 & 0.6 \\ 0.4 & 0.3 & -0.9 & 0.2 \\ 0.2 & 0.9 & 0.3 & -1.4 \end{pmatrix}$$

## Alarm Clock Description

- If the current state is i, the time to the next event is exponentially distributed with rate  $-q_{ii}$  defined to be  $q_i$ .
- Given a transition occurs from state i, the probability that the transition is to state j is proportional to  $q_{ij}$ , namely  $q_{ij}/\sum_{k\neq i}q_{ik}$ .

#### Transition Probabilities

- For a continuous time Markov chain, the *transition matrix* whose *ij* element is the probability of being in state *j* at time *t* given the process begins in state *i* at time 0 is  $P(t) = e^{Qt}$ .
- A probability transition matrix has non-negative values and each row sums to one.
- Each row contains the probabilities from a probability distribution on the possible states of the Markov process.

### **Examples**

$$P(0.1) = \begin{pmatrix} 0.897 & 0.029 & 0.055 & 0.019 \\ 0.019 & 0.899 & 0.029 & 0.053 \\ 0.037 & 0.029 & 0.916 & 0.019 \\ 0.019 & 0.080 & 0.029 & 0.872 \end{pmatrix} \qquad P(0.5) = \begin{pmatrix} 0.605 & 0.118 & 0.199 & 0.079 \\ 0.079 & 0.629 & 0.118 & 0.174 \\ 0.132 & 0.118 & 0.671 & 0.079 \\ 0.079 & 0.261 & 0.118 & 0.542 \end{pmatrix}$$

$$P(1) = \begin{pmatrix} 0.407 & 0.190 & 0.276 & 0.126 \\ 0.126 & 0.464 & 0.190 & 0.219 \\ 0.184 & 0.190 & 0.500 & 0.126 \\ 0.126 & 0.329 & 0.190 & 0.355 \end{pmatrix} \qquad P(10) = \begin{pmatrix} 0.200 & 0.300 & 0.300 & 0.200 \\ 0.200 & 0.300 & 0.300 & 0.300 \\ 0.200 & 0.300 & 0.300 & 0.200 \\ 0.200 & 0.300 & 0.300 & 0.200 \\ 0.200 & 0.300 & 0.300 & 0.200 \end{pmatrix}$$

## The Stationary Distribution

- Well behaved continuous-time Markov chains have a *stationary* distribution, often designated  $\pi$  (not the constant close to 3.14 related to circles).
- When the time t is large enough, the probability  $P_{ij}(t)$  will be close to  $\pi_j$  for each i. (See P(10) from earlier.)
- The stationary distribution can be thought of as a long-run average—over a long time, the proportion of time the state spends in state i converges to  $\pi_i$ .

### **Parameterization**

• The matrix  $Q = \{q_{ij}\}$  is typically parameterized as  $q_{ij} = r_{ij}\pi_j/\mu$  for  $i \neq j$  which guarantees that  $\pi$  will be the stationary distribution when  $r_{ij} = r_{ji}$ .

# Scaling

• The expected number of substitutions per unit time is the average rate of substitution which is a weighted average of the rates for each state weighted by their stationary distribution.

$$\mu = \sum_{i} \pi_{i} q_{i}$$

• If the matrix Q is reparameterized so that all elements are divided by  $\mu$ , then the unit of measurement becomes one substitution.

### Time-reversibility

• The matrix Q is the matrix for a time-reversible Markov chain when  $\pi_i q_{ij} = \pi_j q_{ji}$  for all i and j. That is the overall rate of substitutions from i to j equals the overall rate of substitutions from j to i for every pair of states i and j.