Substitution Models

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Markov chains

- similar to hidden Markov models: states and transitions, but no emissions
- more formally, sequence of random variables X_0, X_1, \ldots, X_n such that state at time t depends only on the state at time t-1 and not on previous states:

$$P(X_t|X_0,\ldots,X_{t-1}) = P(X_t|X_{t-1})$$

- ullet Homogeneous Markov chains: $P(X_t|X_{t-1})$ does not depend on t
- Transition probability matrix: probabilities of moving from one state to another

$$M[x,y] = P(X_t = y | X_{t-1} = x)$$

Example: states {A, C, G, T} can be used to track mutations at a specific position in a chromosome at a specific time point

Markov chains - Stationary distribution (/equilibrium)

- fundamental concept that describes the long-term behavior of a chain
- ullet A distribution π over the set of states is called stationary for a Markov chain with transition matrix M if for every y it holds that

$$\sum_{x} \pi(x) M[x, y] = \pi(y)$$

(or in matrix notation $\pi M = \pi$)

- ullet after many steps $(t o \infty)$, all rows of the matrix converge to the stationary distribution
- A Markov chain can:
 - start in a stationary distribution and therefore remain in it
 - approach the equilibrium as time passes (speed depends on chain structure)
- A chain always converges (regardless of the initial state) if it is ergodic:

 M_t for some t>0 has all entries nonzero

Substitution models, notation

 $P(b | \boldsymbol{a}, t)$: probability that if we start with symbol \boldsymbol{a} , after time t we will see symbol b

Substitution models, notation

P(b|a,t): probability that if we start with symbol a, after time t we will see symbol b

Transition probability matrix:

$$S(t) = \begin{pmatrix} P(A|A,t) & P(C|A,t) & P(G|A,t) & P(T|A,t) \\ P(A|C,t) & P(C|C,t) & P(G|C,t) & P(T|C,t) \\ P(A|G,t) & P(C|G,t) & P(G|G,t) & P(T|G,t) \\ P(A|T,t) & P(C|T,t) & P(G|T,t) & P(T|T,t) \end{pmatrix}$$

Substitution models, basic properties

• S(0) = I

$$\bullet \lim_{t \to \infty} S(t) = \begin{pmatrix} \pi_A & \pi_C & \pi_G & \pi_T \\ \pi_A & \pi_C & \pi_G & \pi_T \\ \pi_A & \pi_C & \pi_G & \pi_T \\ \pi_A & \pi_C & \pi_G & \pi_T \end{pmatrix}$$

Distribution π is called stationary (equilibrium)

- $S(t_1 + t_2) = S(t_1)S(t_2)$ (multiplicativity)
- Jukes-Cantor model should also satisfy

$$S(t) = \begin{pmatrix} 1 - 3s(t) & s(t) & s(t) & s(t) \\ s(t) & 1 - 3s(t) & s(t) & s(t) \\ s(t) & s(t) & 1 - 3s(t) & s(t) \\ s(t) & s(t) & s(t) & 1 - 3s(t) \end{pmatrix}$$

$$S(t) = \begin{pmatrix} 1 - 3s(t) & s(t) & s(t) & s(t) \\ s(t) & 1 - 3s(t) & s(t) & s(t) \\ s(t) & s(t) & 1 - 3s(t) & s(t) \\ s(t) & s(t) & s(t) & 1 - 3s(t) \end{pmatrix}$$

$$S(\epsilon) = \begin{pmatrix} 1 - 3s(\epsilon) & s(\epsilon) & s(\epsilon) & s(\epsilon) \\ s(\epsilon) & 1 - 3s(\epsilon) & s(\epsilon) & s(\epsilon) \\ s(\epsilon) & s(\epsilon) & 1 - 3s(\epsilon) & s(\epsilon) \\ s(\epsilon) & s(\epsilon) & s(\epsilon) & 1 - 3s(\epsilon) \end{pmatrix}$$

Jukes-Cantor model

$$S(t) = \begin{pmatrix} (1+3e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 \\ (1-e^{-4\alpha t})/4 & (1+3e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 \\ (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1+3e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 \\ (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1+3e^{-4\alpha t})/4 \end{pmatrix}$$

Equilibrium?

$$S(t) = \begin{pmatrix} 1 - 3s(t) & s(t) & s(t) & s(t) \\ s(t) & 1 - 3s(t) & s(t) & s(t) \\ s(t) & s(t) & 1 - 3s(t) & s(t) \\ s(t) & s(t) & s(t) & 1 - 3s(t) \end{pmatrix}$$

$$S(2t) = S(t)^2 =$$

$$= \begin{pmatrix} 1 - 6s(t) + 12s(t)^2 & 2s(t) - 4s(t)^2 & 2s(t) - 4s(t)^2 & 2s(t) - 4s(t)^2 \\ 2s(t) - 4s(t)^2 & 1 - 6s(t) + 12s(t)^2 & 2s(t) - 4s(t)^2 & 2s(t) - 4s(t)^2 \\ 2s(t) - 4s(t)^2 & 2s(t) - 4s(t)^2 & 1 - 6s(t) + 12s(t)^2 & 2s(t) - 4s(t)^2 \\ 2s(t) - 4s(t)^2 & 2s(t) - 4s(t)^2 & 2s(t) - 4s(t)^2 & 1 - 6s(t) + 12s(t)^2 \end{pmatrix}$$

$$\approx \left(\begin{array}{cccc} 1 - 6s(t) & 2s(t) & 2s(t) & 2s(t) \\ 2s(t) & 1 - 6s(t) & 2s(t) & 2s(t) \\ 2s(t) & 2s(t) & 1 - 6s(t) & 2s(t) \\ 2s(t) & 2s(t) & 2s(t) & 1 - 6s(t) \end{array} \right)$$
 for $t \to 0$

Substitution rate matrix (matica rýchlostí, matica intenzít)

Substitution rate matrix for Jukes-Cantor model:

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

- ullet For very small t we have S(t) pprox I + Rt
- ullet Rate lpha is the probablity of a change per unit of time for very small t, or derivative of s(t) with respect to t at t=0
- Solving the differential equation for the Jukes-Cantor model we get $s(t) = (1-e^{-4\alpha t})/4$

$$S(t) = \begin{pmatrix} 1 - 3s(t) & s(t) & s(t) & s(t) \\ s(t) & 1 - 3s(t) & s(t) & s(t) \\ s(t) & s(t) & 1 - 3s(t) & s(t) \\ s(t) & s(t) & s(t) & 1 - 3s(t) \end{pmatrix}$$

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

Jukes-Cantor model

$$S(t) = \begin{pmatrix} (1+3e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 \\ (1-e^{-4\alpha t})/4 & (1+3e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 \\ (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1+3e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 \\ (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1-e^{-4\alpha t})/4 & (1+3e^{-4\alpha t})/4 \end{pmatrix}$$

- The rate matrix is typically normalized so that there is on average one substitution per unit of time (time step = substitution occurred)
- In Jukes-Cantor model, $\alpha=1/3$: it represents the equal probability of transition to any of the three other nucleotides from the current nucleotide

Jukes-Cantor model, summary

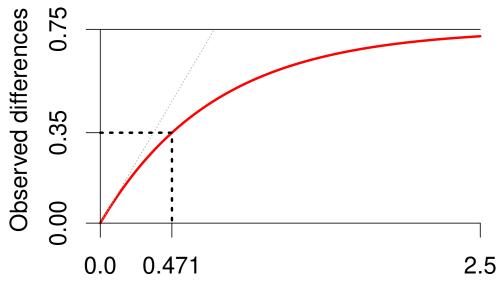
- S(t): matrix 4×4 , where $S(t)_{a,b} = P(b|a,t)$ is the probability that if we start with base a, after time t we have base b.
- ullet Jukes-Cantor model assumes that P(b|a,t) is the same for all a
 eq b
- ullet For a given time t, off-diagonal elements are s(t), diagonal 1-3s(t)
- Rate matrix R: for J-C off-diagonal α , diagonal -3α
- ullet For very small t we have S(t) pprox I Rt
- Rate α is the probability of a change per unit of time for very small t, or derivative of s(t) with respect to t for t=0
- Solving the differential equation for the Jukes-Cantor model, we get $s(t)=(1-e^{-4\alpha t})/4$
- The rate matrix is typically normalized so that there is on average one substitution per unit of time, that is, $\alpha=1/3$

Correction of evolutionary distances

$$\Pr(X_{t_0+t} = C \mid X_{t_0} = A) = \frac{1}{4}(1 - e^{-\frac{4}{3}t})$$

The expected number of observed changes per base in time t:

$$D(t) = \Pr(X_{t_0+t} \neq X_{t_0}) = \frac{3}{4}(1 - e^{-\frac{4}{3}t})$$



Branch length (time)

Correction of observed distances

$$D = \frac{3}{4} \left(1 - e^{-\frac{4}{3}t} \right) \qquad \Rightarrow \qquad t = -\frac{3}{4} \ln \left(1 - \frac{4}{3}D \right)$$

More complex models

ullet General rate matrix R

$$R = \begin{pmatrix} . & \mu_{AC} & \mu_{AG} & \mu_{AT} \\ \mu_{CA} & . & \mu_{CG} & \mu_{CT} \\ \mu_{GA} & \mu_{GC} & . & \mu_{GT} \\ \mu_{TA} & \mu_{TC} & \mu_{TG} & . \end{pmatrix}$$

- ullet μ_{xy} is the rate at which base x changes to a different base y
- ullet Namely, $\mu_{xy} = \lim_{t o 0} rac{\Pr(y \,|\, x, t)}{t}$
- The diagonal is added so that the sum of each row is 0
- General case: 12 parameters. Jukes-Cantor: 1 parameter (α)
- There are models with a smaller number of parameters (compromise between Jukes-Cantor and an arbitrary matrix)

Kimura model

- A and G are purines, C and T pyrimidines
- Purines more often change to other purines and pyrimidines to pyrimidines
- Transition: change within group $A \Leftrightarrow G, C \Leftrightarrow T$,
 Transversion: change to a different group $\{A,G\} \Leftrightarrow \{C,T\}$
- Two parameters: rate of transitions α , rate of transversions β

$$\bullet R = \begin{pmatrix} -2\beta - \alpha & \beta & \alpha & \beta \\ \beta & -2\beta - \alpha & \beta & \alpha \\ \alpha & \beta & -2\beta - \alpha & \beta \\ \beta & \alpha & \beta & -2\beta - \alpha \end{pmatrix}$$

HKY model (Hasegawa, Kishino, Yano)

- Extension of Kimura model, which allows different probabilities of A, C, G, T in the equilibrium
- If we set time to infinity, original base is not important, base frequencies stabilize in an equilibrium.
- Jukes-Cantor has probability of each base in the equilibrium 1/4.
- In HKY the equilibrium frequencies $\pi_A, \pi_C, \pi_G, \pi_T$ are parameters (summing to 1)
- Parameter κ : transition / transversion ratio (α/β)
- $\bullet \ \ \text{Rate matrix:} \ \mu_{x,y} = \left\{ \begin{array}{ll} \kappa \pi_y & \text{if mutation from } x \text{ to } y \text{ is transition} \\ \pi_y & \text{if mutation from } x \text{ to } y \text{ is transversion} \end{array} \right.$

From rate matrix R to transition probabilities S(t)

- ullet J-C and some other models have explicit formulas for S(t)
- For more complex models, such formulas are not available
- In general, $S(t) = e^{Rt}$
- Exponential of a matrix A is defined as $e^A = \sum_{k=0}^{\infty} \frac{1}{k!} A^k$
- If R is diagonalized $R = UDU^{-1}$, where
 - -D is a diagonal matrix with eigenvalues
 - $-\ U$ is the matrix of eigenvectors
- \bullet then $e^{Rt}=Ue^{Dt}U^{-1}$ and the exponential function is applied to the diagnal elements of D (more efficient)
- ullet Note: diagonalization always exists for symmetric matrices R