

Evolution and sequence similarity

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6. marca, 2018

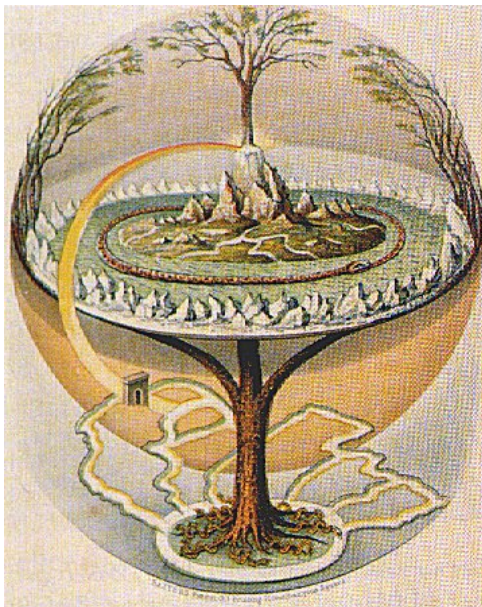
Tree of life?

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Evolution of
DNA

Protein world



Molecular tree of life

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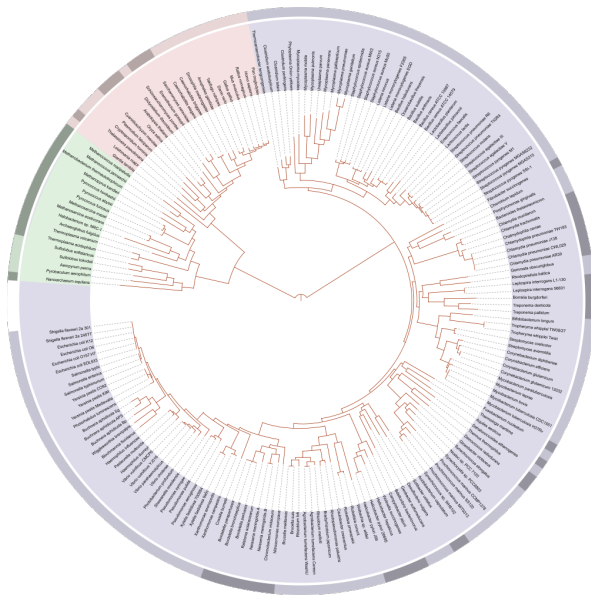
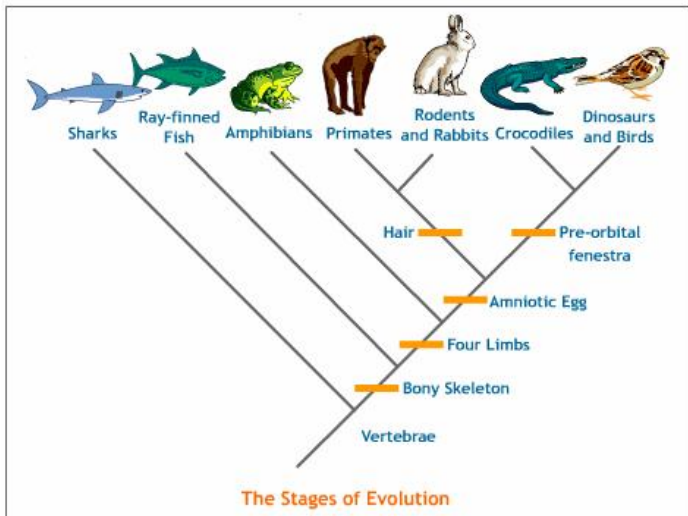


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We are not the “most” evolved species

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DNA replication enables inheritance

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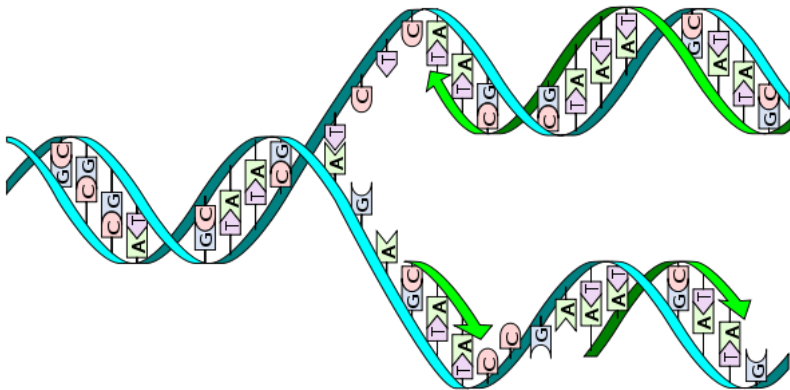
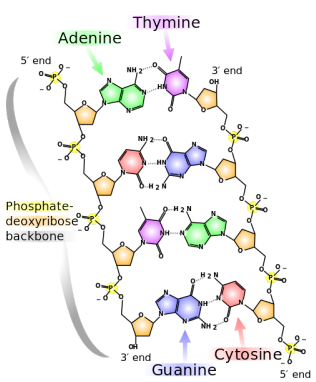
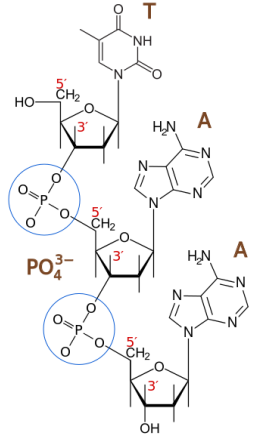


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DNA replication – mechanism

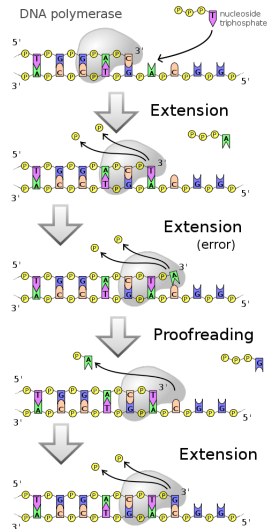


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- DNA polymerase is the key enzyme for DNA replication
- During replication, helper enzymes carry out “proof-reading” of the replicated strand
- error rate (under no stress) $< 10^{-7}$ nucleotides

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GTCTGTAGTA

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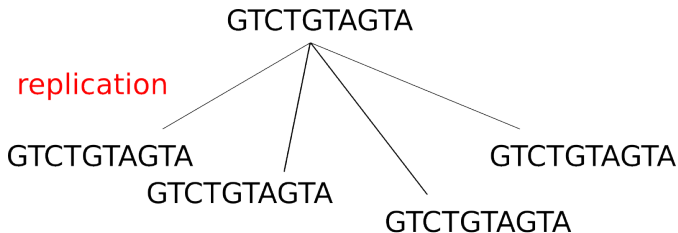


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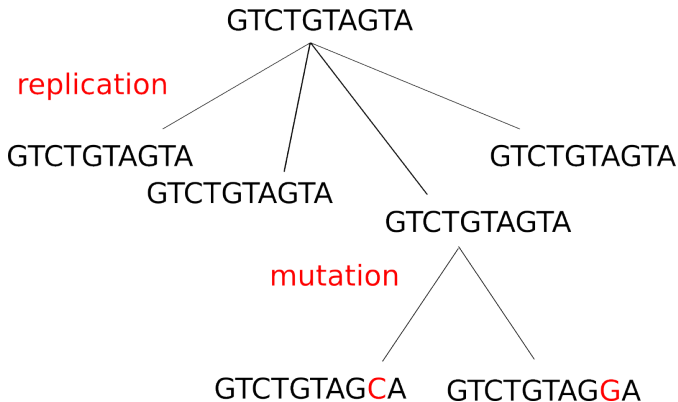


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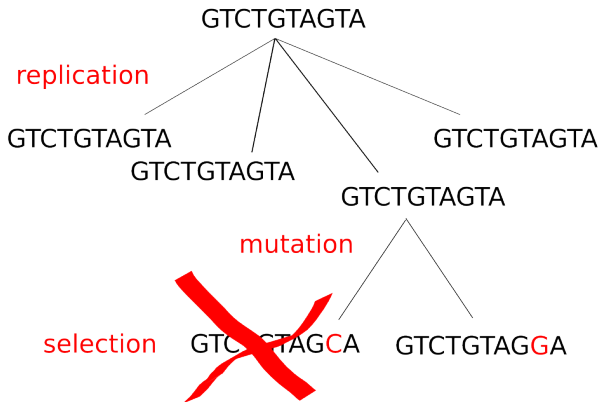


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- How far in evolution are sequences we can observe in different living species?
- More formally: Can we define a measure of sequence similarity

$$d : \Sigma^* \times \Sigma^* \rightarrow \mathcal{R}^+$$

approximating the true evolutionary distance?

- Hint: We should count the number of mutations leading to the observed divergence.

Subproblem 1: multiple scenarios

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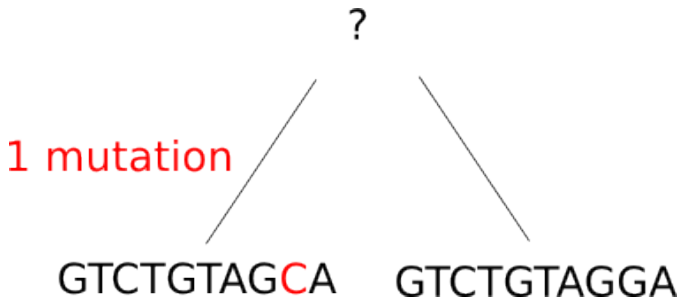
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We can observe only the current situation. What about ancestral sequences?



Solution: *Parsimony* – In case of lack of evidence for a more complex situation, take the simplest possible explanation.

Subproblem 2: Time reversibility



Technically, in order to estimate the ancestral sequence, we need to assume that the process is “time-reversible”, i.e. In the stable state, the rates of mutating the sequence s_1 into s_2 are the same as s_2 into s_1 . This is a reasonable simplification for “short” evolutionary time-scales.

- Time-reversible Markov Chain
- Sequences from Σ^k are states (How many of them?)
- Transition probabilities assume independent base substitution
- We need to define a symmetric base *substitution matrix*
- (*) In fact, we should consider a continuous-time Markov chain, to avoid problems with exact generation times...

The simplest model JC69 (Jukes-Cantor, 1969)

Only one parameter: μ

$$Q = \begin{pmatrix} * & \frac{\mu}{4} & \frac{\mu}{4} & \frac{\mu}{4} \\ \frac{\mu}{4} & * & \frac{\mu}{4} & \frac{\mu}{4} \\ \frac{\mu}{4} & \frac{\mu}{4} & * & \frac{\mu}{4} \\ \frac{\mu}{4} & \frac{\mu}{4} & \frac{\mu}{4} & * \end{pmatrix}$$

Solution for continuous time t :

$$P = \begin{pmatrix} \frac{1}{4} + \frac{3}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} \\ \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} + \frac{3}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} \\ \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} + \frac{3}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} \\ \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} - \frac{1}{4}e^{-t\mu} & \frac{1}{4} + \frac{3}{4}e^{-t\mu} \end{pmatrix}$$

Kimura 1980 (K80) model

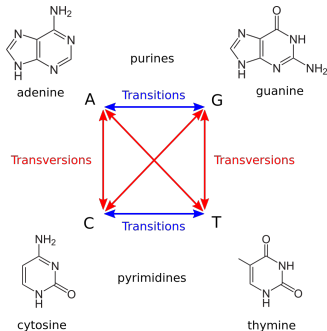


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- We can observe that transitions are different than transversions. This leads to the Kimura model (with p, q being the probability of transition, transversion).

$$Q = \begin{pmatrix} * & \kappa & 1 & 1 \\ \kappa & * & 1 & 1 \\ 1 & 1 & * & \kappa \\ 1 & 1 & \kappa & * \end{pmatrix}$$

$$K = -\frac{1}{2} \ln((1 - 2p - q)\sqrt{1 - 2q})$$

Felsenstein model F81 (Felsenstein, 1981)

We do not assume equal probability of nucleotides, but a distribution, with

$$\pi_A \neq \pi_C \neq \pi_G \neq \pi_T$$

Then the mutation rate matrix may look like the following

$$Q = \begin{pmatrix} * & \pi_G & \pi_C & \pi_T \\ \pi_A & * & \pi_C & \pi_T \\ \pi_A & \pi_G & * & \pi_T \\ \pi_A & \pi_G & \pi_C & * \end{pmatrix}$$

- Mutations occur on DNA level, but selection acts much higher: on the phenotype level.
- This makes the assumption of base independence invalid
- Long evolutionary times violate time-reversibility
- Multiplicative measure not too convenient in practice
- We can only account for substitutions, not for insertions or deletions

Suggested solutions:

- Use protein sequences for comparisons
- Define additive substitution matrices

mRNA translation into proteins

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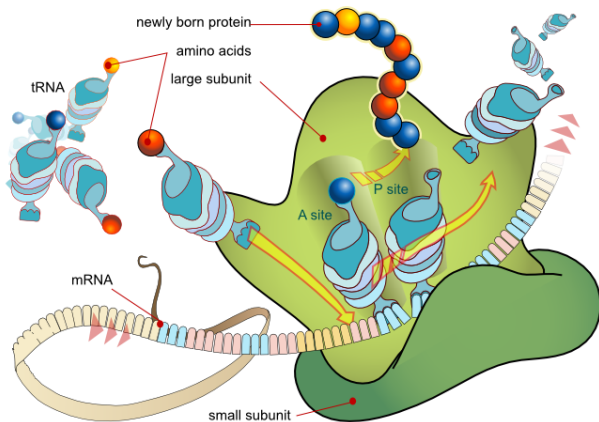


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Protein codon table

		Second base					
		U	C	A	G		
First base	U	UUU } Phenyl- UUC } alanine F	UCU } UCC } Serine S	UAU } Tyrosine Y UAC } UAA } Stop codon UAG } Stop codon	UGU } Cysteine C UGC } UGA } Stop codon UGG } Tryptophan W	U	C
	C	CUU } CUC } Leucine L	CCU } CCC } Proline P	CAU } Histidine H CAC } CAA } Glutamine Q CAG }	CGU } CGC } Arginine R CGA } CGG }	C	A
	A	AUU } Isoleucine I AUC } AUA } AUG } Methionine start codon M	ACU } ACC } Threonine T ACA } ACG }	AAU } Asparagine N AAC } AAA } Lysine K AAG }	AGU } Serine S AGC } AGA } Arginine R AGG }	A	G
	G	GUU } GUC } Valine V GUA } GUG }	GCU } GCC } Alanine A GCA } GCG }	GAU } Aspartic acid D GAC } GAA } Glutamic acid E GAG }	GGU } GGC } Glycine G GGA } GGG }	G	U
						C	A
						G	U
						C	A
						G	U
						C	A
						G	U
						C	A
						G	U

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- We are still assuming time-reversible Markov chain, but now in space of protein sequences.
- Matrix entries contain log-probabilities, leading to additive measures of similarity
- PAM (Point accepted mutations) matrices (Dayhoff, 1978) describe observed probabilities of occurrence of point mutations for a given average divergence (PAM1 = one mutation/100 bases, mostly used PAM250)
- BLOSUM (BLOcks Substitution Matrix) (Henikoff, Henikoff 1992) were constructed using short protein alignments (Blocks) of given sequence identity.
e.g. BLOSUM80 was derived from sequences of $\geq 80\%$ identity