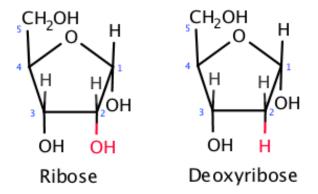
Course Overview

- 1. Biological Background
- 2. Pairwise sequence alignment algorithms
- 3. Probabilistic alignments: Hidden Markov models
- 4. Multiple sequence alignments
- 5. Phylogeny: Algorithms for reconstructing pedigrees
- 6. Recent advances & applications: Inferring Genetic Diversity from Next-generation Sequencing data.
- 7. Neural nets & deep learning for sequence analysis: CNNs, RNNs, Transformers, Alphafold etc.

Short historical Introduction

- Genetics as a natural science started in 1866: Gregor Mendel performed experiments that pointed to the existence of biological elements called genes.
- Deoxy-ribonucleic acid (DNA) isolated by Friedrich Miescher in 1869.
- 1944: Oswald Avery (and coworkers) identified DNA as the major carrier of genetic material, responsible for inheritance.
 Ribose: (simple) sugar molecule, deoxy-ribose ~> loss of oxygen atom.
 Nucleic acid: overall name for DNA and RNA (large biomolecules). Named for their initial discovery in nucleus of cells, and for presence of phosphate groups (related to phosphoric acid).

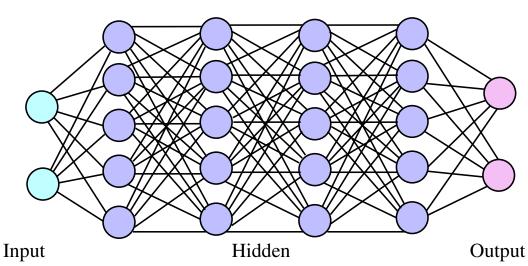


By Miranda19983 - Own work, CC BY-SA 4.0, https://commons.wikimedia.org/w/index.php?curid=84120486

Short historical Introduction

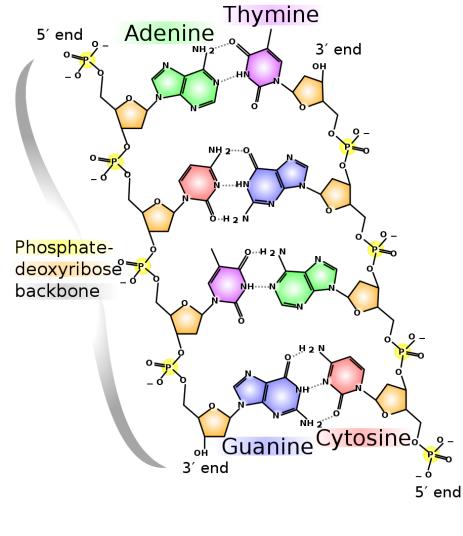
- 1953, Watson & Crick: 3-dimensional structure of DNA. They inferred the method of DNA replication.
- 2001: first draft of the human genome published by the Human Genome Project and the company Celera.
- Many new developments, such as Next Generation Sequencing, Deep learning etc.





By RE73 - Own work, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=18862884

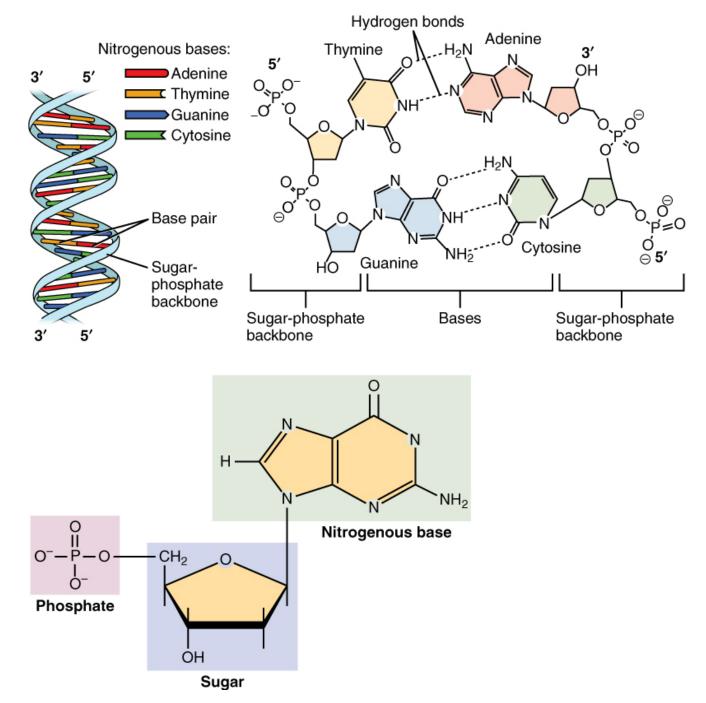
Base pairs and the DNA



By Madprime (talk · contribs) - Own work, CC BY-SA 3.0,

https://commons.wikimedia.org/w/index.php?curid=1848174

- DNA composed of 4 basic molecules
 ~~ nucleotides.
- Nucleotides are identical up to different **nitrogen base:** organic molecule with a nitrogen atom that has the chemical properties of a base (due to free electron pair at nitrogen atom).
- Each nucleotide contains phosphate, sugar (of deoxy-ribose type), and one of the 4 bases: Adenine, Guanine, Cytosine, Thymine (A,G,C,T).
- Hydrogen bonds between base pairs $G \equiv C, A = T.$

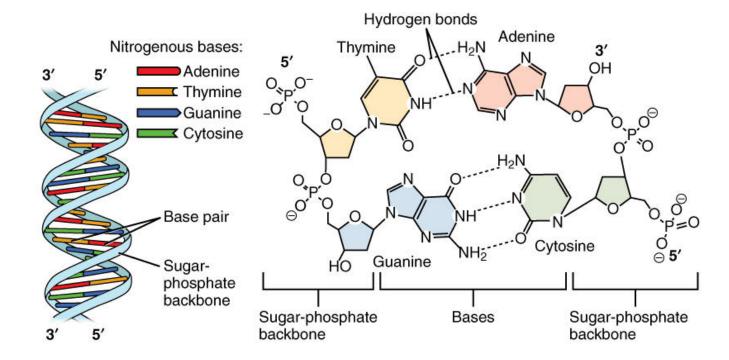


By OpenStax - https://cnx.org/contents/FPtK1zmh@8.25:fEI3C8Ot@10/Preface, CC BY 4.0,

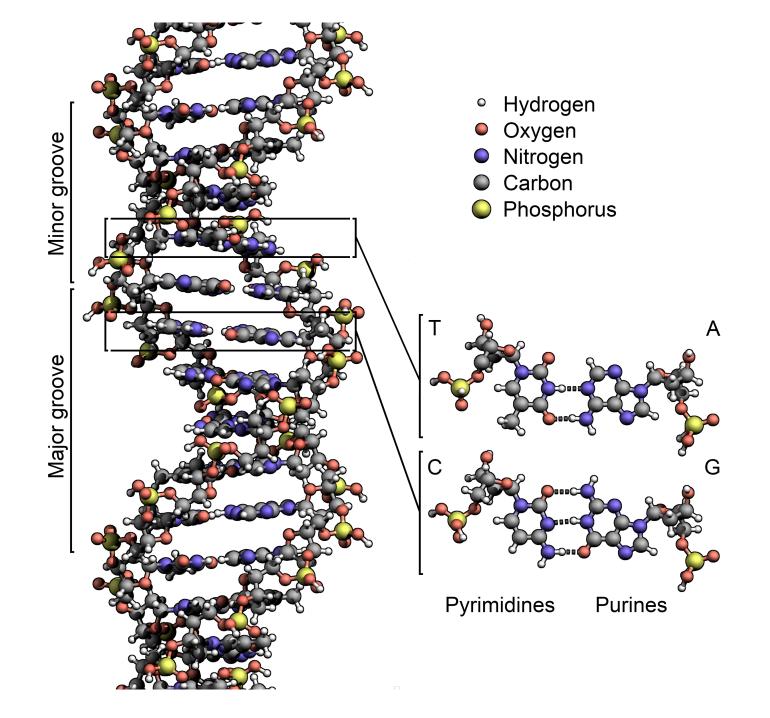
https://commons.wikimedia.org/w/index.php?curid=30131206

The structure of DNA

- DNA molecule is directional due to asymmetrical structure of the sugars which constitute the skeleton: Each sugar is connected to the strand upstream in its 5th carbon and to the strand downstream in its 3rd carbon.
- DNA strand goes from 5' to 3'. The directions of the two complementary DNA strands are reversed to one another (→ Reversed Complement).



Adapted from https://commons.wikimedia.org/w/index.php?curid=30131206



By Zephyris - Own work, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=15027555

Replication of DNA

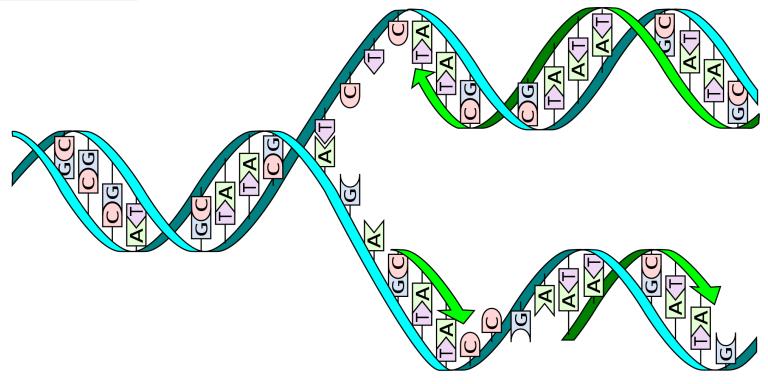
Biological process of producing two replicas of DNA from one original DNA molecule. Cells have the distinctive property of division

→ DNA replication is most essential part for **biological inheritance.**

Unwinding ~>>>>>>>> single bases exposed on each strand.

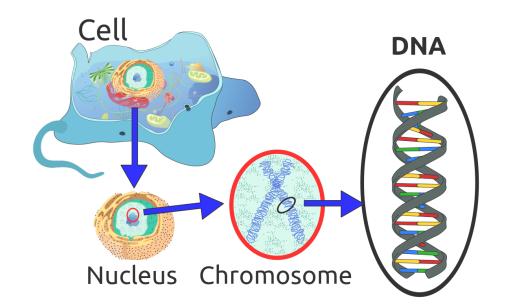
Pairing requirements are **strict** \rightsquigarrow single strands are templates for re-forming **identical** double helix (up to **mutations**).

DNA polymerase: enzyme that catalyzes the synthesis of new DNA.



Genes and Chromosomes

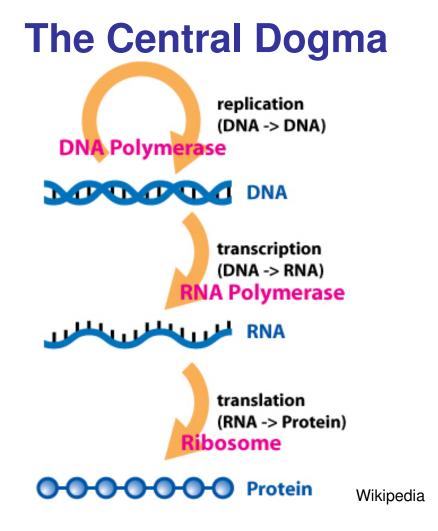
- In higher organisms, DNA molecules are packed in a chromosome.
- **Genome:** total genetic information stored in the chromosomes.
- Every cell contains a **complete set** of the genome, differences are due to variable **expression** of genes.
- A gene is a sequence of nucleotides that encodes the synthesis of a gene product.



By Sponk, Tryphon, Magnus Manske,

https://commons.wikimedia.org/w/index.php?curid=20539140

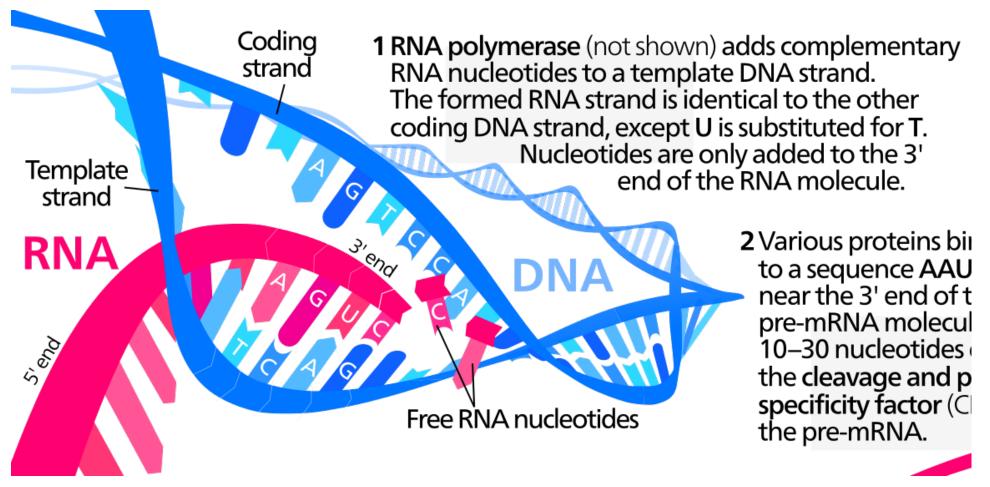
Gene expression: Process of synthesizing a gene product (often a protein)
 ~> controls timing, location, and amount.



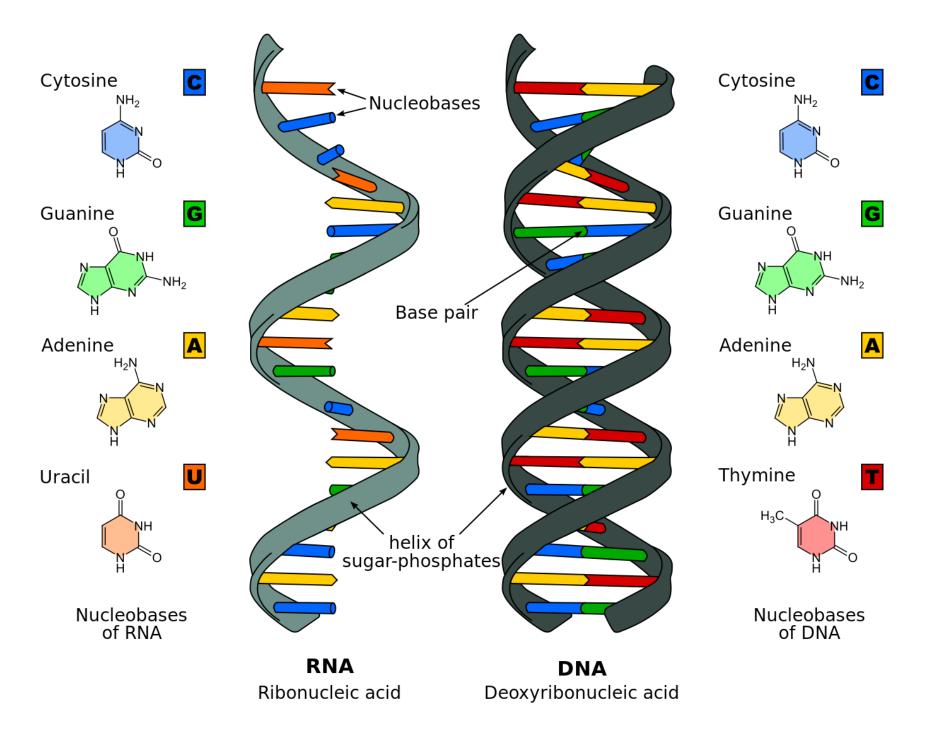
Transcription: making of an RNA molecule from DNA template. **Translation:** construction of amino acid sequence from RNA.

Almost no exceptions (~> retroviruses)

Transcription



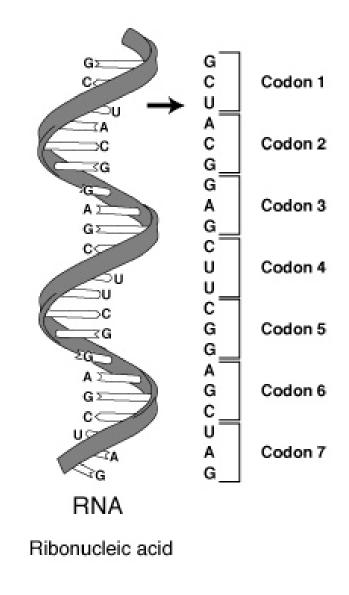
By Kelvinsong - Own work, CC BY 3.0, https://commons.wikimedia.org/w/index.php?curid=23086203

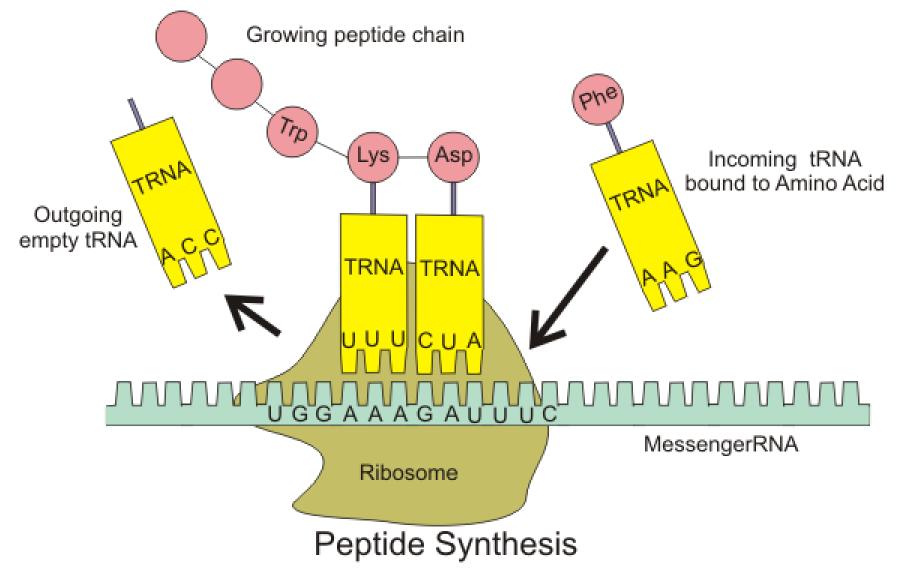


https://commons.wikimedia.org/w/index.php?curid=9810855

Translation

- mRNA molecules are translated by **ribosomes**: Enzyme that links together amino acids.
- Message is read three bases at a time.
- Initiated by the first AUG codon (codon = nucleotide triplet).
- Covalent bonds (=sharing of electron pairs) are made between adjacent amino acids
 ⇒ growing chain of amino acids ("polypeptide").
- When a "stop" codon (UAA, UGA, UAG) is encountered, translation stops.





By Boumphreyfr - Own work, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=7200200

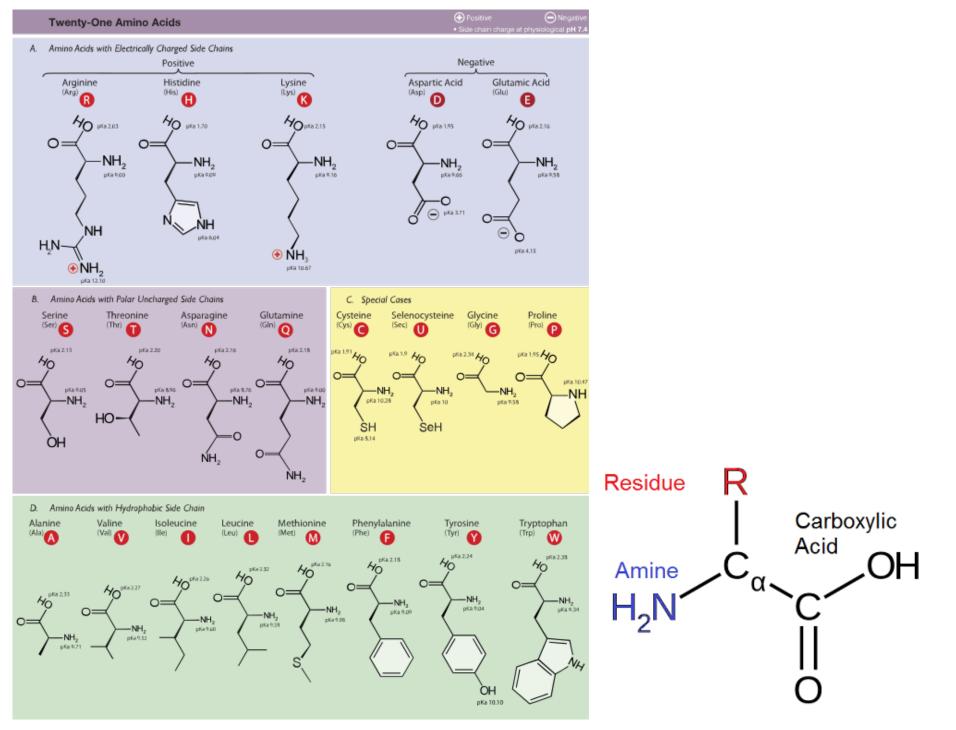
The genetic code

Standard genetic code

1st				2nd	base				3rd
base		U		с		A		base	
U	UUU	(Pho/E) Phosydalapina	UCU		UAU	(Tur M) Turanina	UGU		U
	UUC	(Phe/F) Phenylalanine	UCC	(CastC) Casina	UAC	(Tyr/Y) Tyrosine	UGC	(Cys/C) Cysteine	С
	UUA		UCA	(Ser/S) Serine	UAA ^[8]	Stop (Ochre)	UGA ^[B]	Stop (Opa/)	Α
	UUG		UCG		UAG ^[8]	Stop (Amber)	UGG	(Trp/W) Tryptophan	G
с	CUU	(Leu/L) Leucine	CCU		CAU	(Listan) Listiding	CGU		U
	CUC		CCC	(Pro (P) Proline	CAC	(His/H) Histidine	CGC	(Arm/D) Arminian	С
	CUA		CCA	(Pro/P) Proline	CAA		CGA	(Arg/R) Arginine	Α
	CUG		CCG		CAG	(GIn/Q) Glutamine	CGG		G
	AUU		ACU		AAU	(Asn/N) Asparagine	AGU	(Cor/C) Corino	U
U C A	AUC	(IIe/I) Isoleucine	ACC	(The C) Theoremise	AAC	(Ashini) Asparagine	AGC	(Ser/S) Serine	С
	AUA		ACA	(Thr/T) Threonine	AAA	(Lum (C) Lumino	AGA	(Arm (D)) Arminian	Α
	AUG ^[A]	(Met/M) Methionine	ACG		AAG	(Lys/K) Lysine	AGG	(Arg/R) Arginine	G
	GUU		GCU		GAU	(Asp(D)) Aspertia sold	GGU		U
~	GUC	0.000	GCC		GAC	(Asp/D) Aspartic acid	GGC		С
G	GUA	(Val/V) Valine	GCA	(Ala/A) Alanine	GAA	(Olu/E) Olutomia anid	GGA	(Gly/G) Glycine	Α
	GUG		GCG		GAG	(Glu/E) Glutamic acid	GGG		G

Wikipedia

Highly redundant: only 20 (or 21) amino acids formed from $4^3 = 64$ possible combinations.



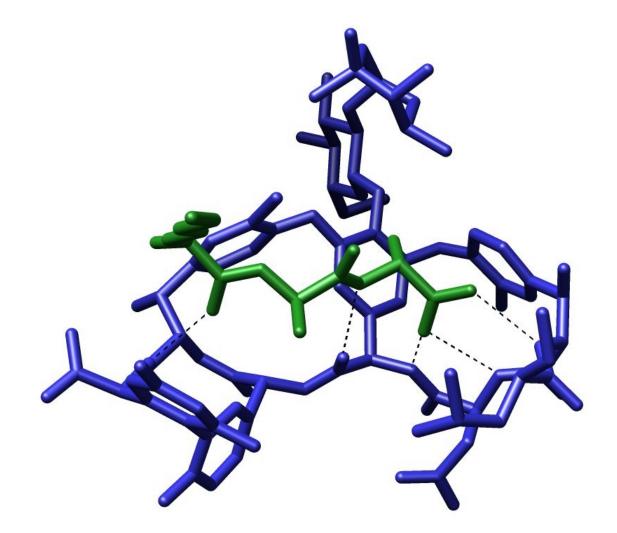
By Dancojocari. https://commons.wikimedia.org/w/index.php?curid=9176441

Proteins

- Linear polymer of amino acids, linked together by peptide bonds. Average size ≈ 200 amino acids, can be over 1000.
- To a large extent, cells are made of proteins.
- Proteins determine shape and structure of a cell.
 Main instruments of molecular recognition and catalysis.
- **Complex structure** with four hierarchical levels.
 - 1. Primary structure: amino acid sequence.
 - 2. Different regions form locally regular secondary structures like α -helices and β -sheets.
 - 3. **Tertiary structure**: packing such structures into one or several 3D *domains*.
 - 4. Several domains arranged in a **quaternary structure**.

Molecular recognition

Interaction between molecules through noncovalent bonding

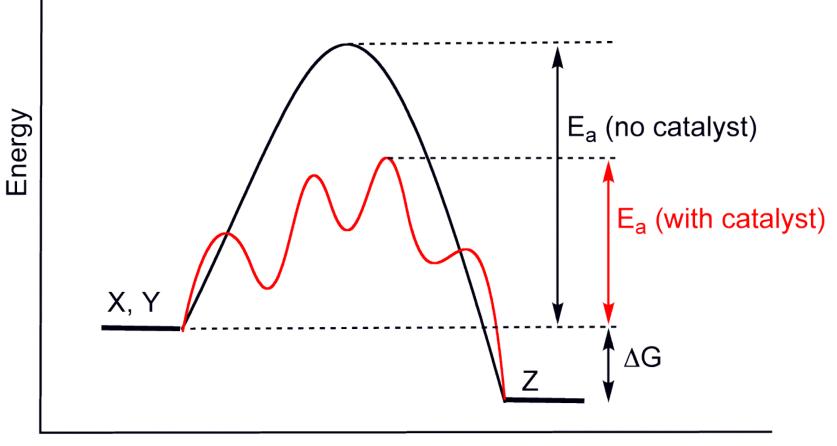


Crystal structure of a short peptide L-Lys-D-Ala-D-Ala (bacterial cell wall precursor) bound to the antibiotic vancomycin through hydrogen

bonds. By M stone, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=2327682

Catalysis

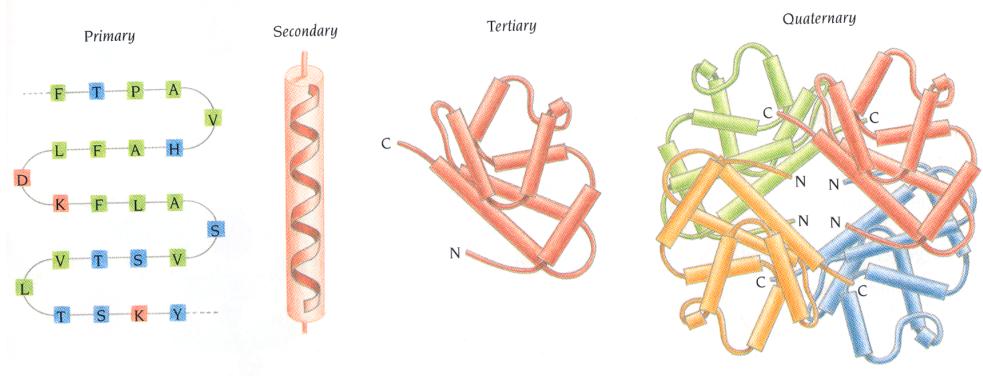
Increasing the rate of a chemical reaction by adding a substance \rightsquigarrow catalyst.



Reaction Progress

Wikipedia

Protein Structure: primary to quaternary



Durbin et al., Cambridge University Press. https://doi.org/10.1017/CBO9780511790492.004

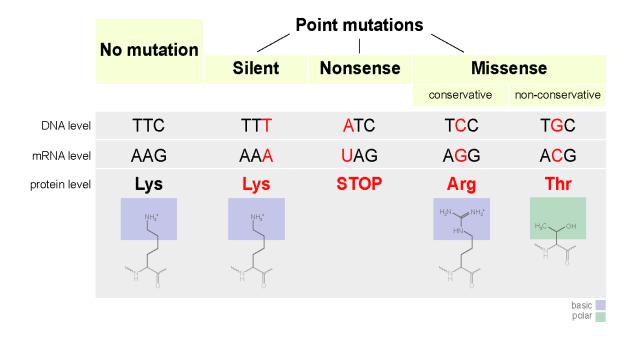
Structure is determined by the **primary sequence** and their **physico-chemical interactions** in the medium. **Structure determines functionality.**

Mutations

- Mutation: Heritable change in the DNA sequence. Occur due to exposure to ultra violet radiation or other environmental conditions.
- **Two levels** at which a mutation can take place:
 - **Point mutation:** within a single gene.
 - substitution (change of one nucleotide),
 - insertion (addition of nucleotides),
 - deletion.
 - Chromosomal mutation: whole segments interchange, either on the same chromosome, or on different ones.

Point Mutations

- May arise from spontaneous mutations during DNA replication.
- The rate of mutation increased by **mutagens** (physical or chemical agent that changes the genetic material).
- Mutagens: Physical (UV-, X-rays or heat), or chemical (molecules misplace base pairs / disrupt helical shape of DNA).



Wikipedia. By Jonsta247 - Own work, CC BY-SA 4.0, https://commons.wikimedia.org/w/index.php?curid=12481467

Importance of Mutations

Mutations are responsible for inherited disorders & diseases.
 Sickle-cell anemia caused by missense point mutation in hemoglobin (in blood cells, responsible for oxygen transport.)
 Hydrophilic glutamic acid replaced with hydrophobic valine.
 ~> deformed red blood cells.

Sequence for Normal Hemoglobin: 6th codon: adenine (A)

AUG	GUG	CAC	CUG	ACU	CCU	G <mark>A</mark> G	GAG	AAG	UCU	GCC	GUU	ACU
START	Val	His	Leu	Thr	Pro	Glu	Glu	Lys	Ser	Ala	Val	Thr

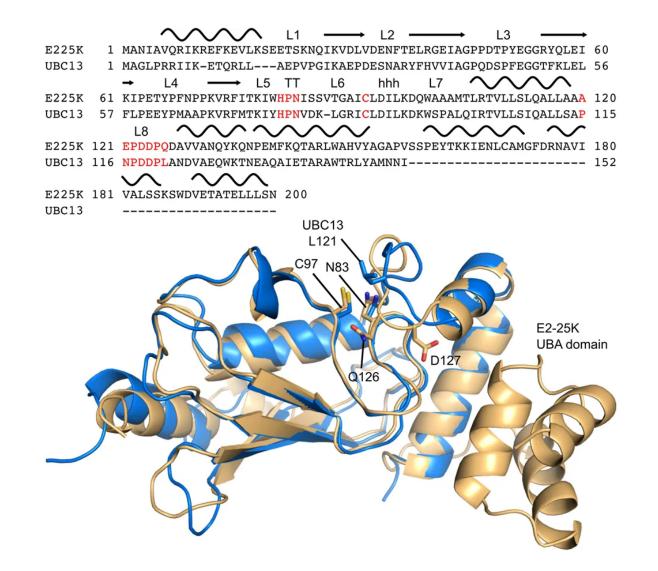
Sickle Cell Hemoglobin: ~> thymine (DNA), uracil (RNA)

AUG	GUG	CAC	CUG	ACU	CCU	G <mark>U</mark> G	GAG	AAG	UCU	GCC	GUU	ACU
START	Val	His	Leu	Thr	Pro	Val	Glu	Lys	Ser	Ala	Val	Thr

- Mutations are the source of phenotypic variation
 - \Rightarrow **new species** and **adaption** to environmental conditions.

Sequence Comparison: Motivation

Basic idea: similar sequences \rightsquigarrow similar proteins. Protein folding: 30 % sequence identity \Rightarrow structures similar.

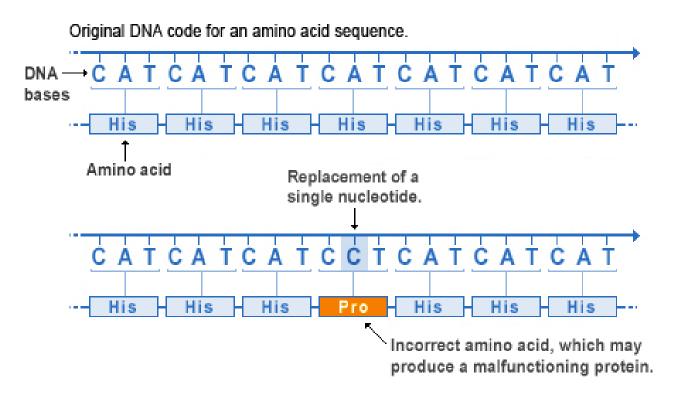


Rout et al., Scientific Reports, vol 8, no 7002 (2018)

Comparing sequences

Theory: during evolution **mutations** occurred, creating differences between families of contemporary species.

Missense mutation



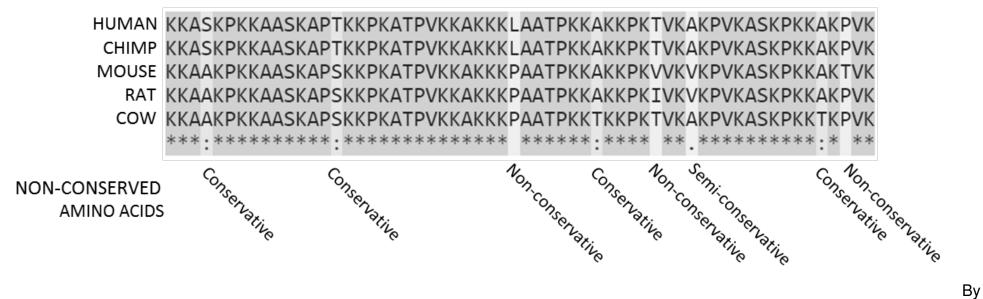
U.S. National Library of Medicine

https://commons.wikimedia.org/w/index.php?curid=25399199

Comparing sequences

Comparing two sequences: looking for **evidence** that they have **diverged from a common ancestor** by a **mutation process**.





Thomas Shafee - Own work, CC BY 4.0, https://commons.wikimedia.org/w/index.php?curid=37188728

Sequence Alignment

Informal definition:

Alignment of sequences $x = x_1 \dots x_n$ and $y = y_1 \dots y_m$:

(i) insert spaces,

(ii) place resulting sequences **one above the other** so that every character or space has a counterpart.

Example: ACBCDDDB and CADBDAD. Possible alignments:

Optimal Alignment

Given: two sequences x and y over alphabet A.

 $\mathcal{A} = \{A, G, C, T\} \text{ (DNA)}$ $\mathcal{A} = \{A, R, N, D, C, Q, E, G, H, I, L, K, M, F, P, S, T, W, Y, V\} \text{ (proteins)}$

Formalizing optimality of an alignment: define

- the costs for substituting a letter by another letter
 ⇒ substitution matrix;
- the costs for **insertion** \Rightarrow **gap penalties**.

The Scoring Model

- Idea: assign a score to each alignment, choose best one.
- Additive scoring scheme: Total score = sum of all scores for pairs of letters + costs for gaps.

Implicit assumption:

Mutations at different sites have occurred **independently**. (In most cases) reasonable for DNA and protein sequences.

- All common algorithms use additive scoring schemes.
- Modeling dependencies is possible, but at the price of significant computational complexities.

Substitution Matrices

• Expectation:

Identities in real alignments are more likely than by chance.

- Derive score for aligned pairs from a probabilistic model.
- Score: relative likelihood that two sequences are evolutionary related as opposed to being unrelated
 - → score = ratio of probabilities.
- First assumption: Ungapped alignment, n = m.
- R: Random model:

Letter a occurs **independently** with some frequency q_a

 \Rightarrow Pr(two sequences) = product of probabilities for each letter:

$$P(x, y|R) = \prod_{i} q_{x_i} \prod_{i} q_{y_i}.$$

Substitution Matrices

• *M* (match): aligned pairs occur with joint probability

$$P(x, y|M) = \prod_{i} p_{x_i y_i}$$

• Ratio ~~ "odds ratio":

$$\frac{P(x, y|M)}{P(x, y|R)} = \prod_{i} \frac{p_{x_i y_i}}{q_{x_i} q_{y_i}}$$

• To arrive at an **additive** scoring system \rightarrow **log-odds ratio**:

$$S = \sum_{i} \log \left(\frac{p_{x_i y_i}}{q_{x_i} q_{y_i}} \right) = \sum_{i} s(x_i, y_i)$$

 s(a, b): log-likelihood ratio of pair (a, b) occurring as an aligned pair as opposed to an unaligned pair → substitution matrix₃₁

BLOSUM62 substitution matrix

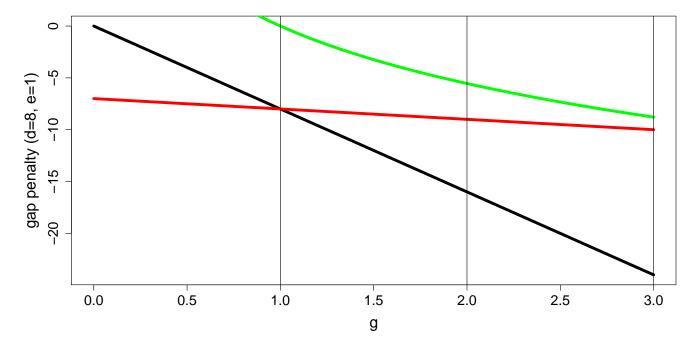
Ala	4																				
Arg	-1	5																			
Asn	-2	0	6																		
Asp	-2	-2	1	6																	
Cys	0	-3	-3	-3	9																
Gln	-1	1	0	0	-3	5															
Glu	-1	0	0	2	-4	2	5														
Gly	0	-2	0	-1	-3	-2	-2	6													
His	-2	0	1	-1	-3	0	0	-2	8												
lle	-1	-3	-3	-3	-1	-3	-3	-4	-3	4											
Leu	-1	-2	-3	-4	-1	-2	-3	-4	-3	2	4										
Lys	-1	2	0	-1	-3	1	1	-2	-1	-3	-2	5									
Met	-1	-1	-2	-3	-1	0	-2	-3	-2	1	2	-1	5								
Phe	-2	-3	-3	-3	-2	-3	-3	-3	-1	0	0	-3	0	6							
Pro	-1	-2	-2	-1	-3	-1	-1	-2	-2	-3	-3	-1	-2	-4	7						
Ser	1	-1	1	0	-1	0	0	0	-1	-2	-2	0	-1	-2	-1	4					
Thr	0	-1	0	-1	-1	-1	-1	-2	-2	-1	-1	-1	-1	-2	-1	1	5				
Trp	-3	-3	-4	-4	-2	-2	-3	-2	-2	-3	-2	-3	-1	1	-4	-3	-2	11			
Tyr	-2	-2	-2	-3	-2	-1	-2	-3	2	-1	-1	-2	-1	3	-3	-2	-2	2	7		
Val	0	-3	-3	-3	-1	-2	-2	-3	-3	3	1	-2	1	-1	-2	-2	0	-3	-1	4	
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His	lle	Leu	Lys	Met	Phe	Pro	Ser	Thr	Trp	Tyr	Val	

Wikipedia

Gap penalties

Gap penalty types for a gap of length g:

- Linear: $\gamma(g) = -gd$, with d being the gap weight.
- Affine: $\gamma(g) = -d (g 1)e$, gap-open penalty d, gap-extension penalty e. Usually e < d.
- Convex: e.g. $\gamma(g) = -d \log(g)$. Each additional space contributes less than the previous space.



Global Alignment: Needleman-Wunsch algorithm

The Global Alignment problem:

INPUT: two sequences $x = x_1 \dots x_n$ and $y = y_1 \dots y_m$.

TASK: Find optimal alignment for linear gap penalties $\gamma(g) = -gd$.

Let F(i, j) be the optimal alignment score of the **prefix sequences** $x_{1...i}$ and $y_{1...j}$. A zero index i = 0 or j = 0 refers to an **empty sequence.** F(i, j) has following properties:

Base conditions:
$$F(i,0) = \sum_{k=1}^{i} -d = -id$$

 $F(0,j) = \sum_{k=1}^{j} -d = -jd, \quad F(0,0) = 0.$

Recurrence relation:

for
$$1 \le i \le n, \ 1 \le j \le m$$
:
 $F(i,j) = \max \begin{cases} F(i-1,j-1) + s(x_i,y_j) \\ F(i-1,j) - d \\ F(i,j-1) - d \end{cases}$

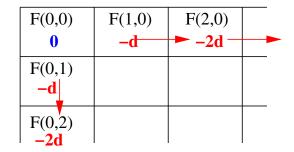
Tabular Computation of Optimal Alignment

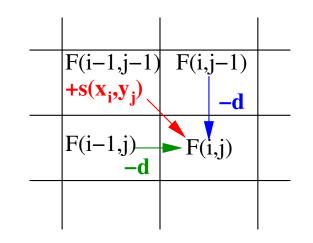
Starting from F(0,0) = 0, fill the whole matrix $(F)_{ij}$:

for i = 0 or j = 0, calculate new value from left-hand (upper) value.

for $i, j \ge 1$, calculate the bottom right-hand corner of each square of 4 cells from one of the 3 other cells:

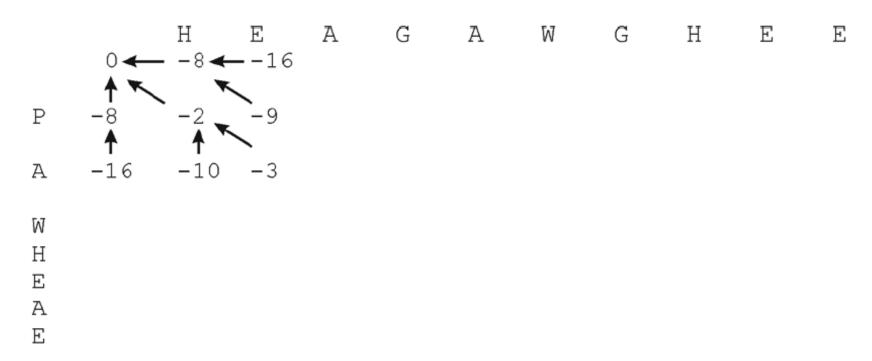
keep a pointer in each cell back to the cell from which it was derived \Rightarrow traceback pointer.





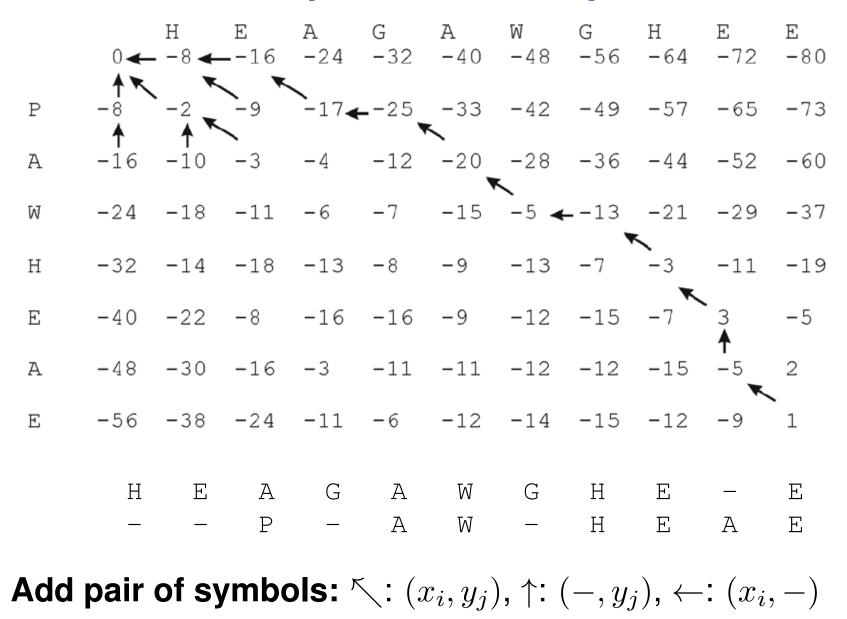
Global Alignment: Example

x = HEAGAWGHEE, y = PAWHEAE. Linear gap costs d = 8. Scoring matrix: BLOSUM50



Durbin et al., Cambridge University Press. https://doi.org/10.1017/CBO9780511790492.004

Example: traceback procedure



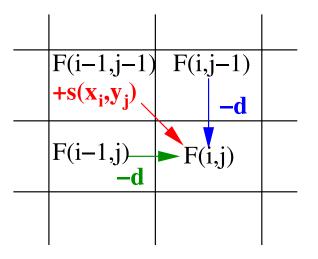
Adapted from Durbin et al., Cambridge University Press. https://doi.org/10.1017/CBO9780511790492.004

Time and Space Complexity

Theorem 1. The time complexity of the Needleman-Wunsch algorithm is O(nm). Space complexity is O(m), if only F(x, y) is required, and O(nm) for the reconstruction of the alignment.

Proof:

Time: when computing F(i, j), only cells (i-1, j-1), (i, j-1), (i-1, j) are examined \rightsquigarrow constant time. There are (n+1)(m+1) cells $\rightsquigarrow O(nm)$ **time complexity.**

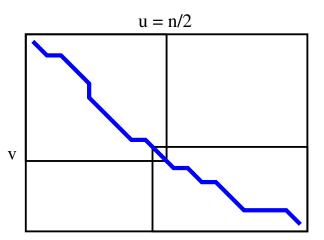


Space : row-wise computation of the matrix: for computing row k, only row k - 1 must be stored $\rightsquigarrow O(m)$ **space. Reconstructing** the alignment: all traceback pointers must be stored $\rightsquigarrow O(nm)$ **space complexity.**

Global Alignment in Linear Space

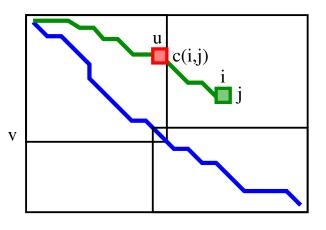
- Problem: genomic scale sequence analysis: comparing two large genomic sequences: $m, n \approx 10^6 \Rightarrow$ space complexity 10^{12} is clearly unacceptable!
- Solution: linear space algorithms with space complexity O(m+n).
- Basic idea: divide and conquer. Let $u = \lfloor \frac{n}{2} \rfloor$ be the integer part of $\frac{n}{2}$.
 - Let v be a row index such that the cell (u, v) is on the optimal alignment.
 - Split dynamic programming problem into two parts: $(0,0) \rightarrow (u,v)$ and $(u,v) \rightarrow (n,m)$. Optimal alignment will be concatenation of individual sub-alignments.

- Repeat splitting until until u = 0: trivial



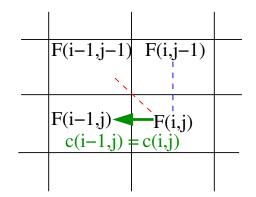
Global Alignment in Linear Space

• For $i \ge u$ define c(i, j) such that (u, c(i, j)) is on the optimal path from $(1, 1) \rightarrow (i, j)$.



 Let (i', j') be the preceding cell to (i, j) from which F(i, j) is derived. Update c(i, j) as:

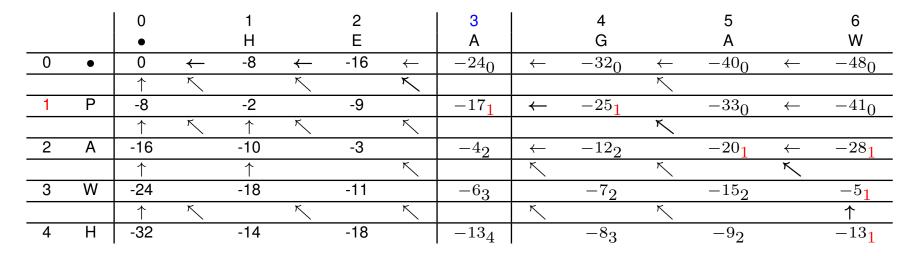
$$c(i,j) = \begin{cases} j & , \text{ if } i = u, \\ c(i',j') & , \text{else} \end{cases}$$



- Local operation \rightsquigarrow need to store only the previous row of c().
- Finally, v = c(n, m).

Global Alignment in Linear Space: Example

Computing the *c* matrix for the first step (i = n = 6, j = m = 4, u = 3). The *c* values are written as subscripts. BLOSUM62, linear gap costs d = 8.



Every c(i, j) defines a row index v such that (u, c(i, j)) is on the optimal path from (1, 1) to $(i, j) \rightsquigarrow v = c(6, 4) = 1$, so (3,1) is our desired element on the optimal path form (1,1) to (6,4).

Local Alignments

The Local Alignment problem:

INPUT: two sequences $x = x_1, \ldots, x_n$ and $y = y_1, \ldots, y_m$.

TASK: find subsequences a of x and b of y,

whose similarity (=optimal global alignment score) is maximal (over all such pairs of subsequences).

Assume linear gap penalties $\gamma(g) = -gd$.

Subsequence = **contiguous** segment of a sequence.

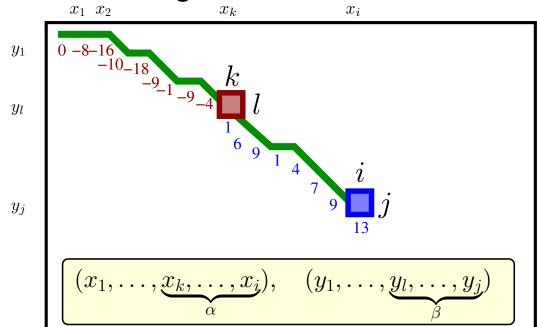
Consider first a simpler problem by fixing the endpoint of the subsequences at index pair (i, j):

Local suffix alignment problem: given x, y, i, j, find suffixes α of $x_{1,...,i}$ and β of $y = y_{1,...,j}$ such that their global alignment score is maximal.

$$(x_1,\ldots,\underbrace{x_k,\ldots,x_i}_{\alpha}), \quad (y_1,\ldots,\underbrace{y_l,\ldots,y_j}_{\beta})$$

Local suffix alignments

Consider global alignment path to cell (i, j). Where to start? Intuition: Indices (k, l) found by following the path back to (0, 0), but stopping at the first negative value.



Remark: If we consider all solutions (i.e. for all (i, j) pairs), we look at all possible subsequences (no restrictions on α, β)

Maximal solution of local suffix alignment over all pairs (i, j) = solution of local alignment problem.

Smith-Waterman Algorithm

F(i, j): optimal local suffix alignment for indices i, j.

Global alignment with one **modification:** Prefixes whose scores are ≤ 0 are **discarded** \rightarrow alignment can **start anywhere**.

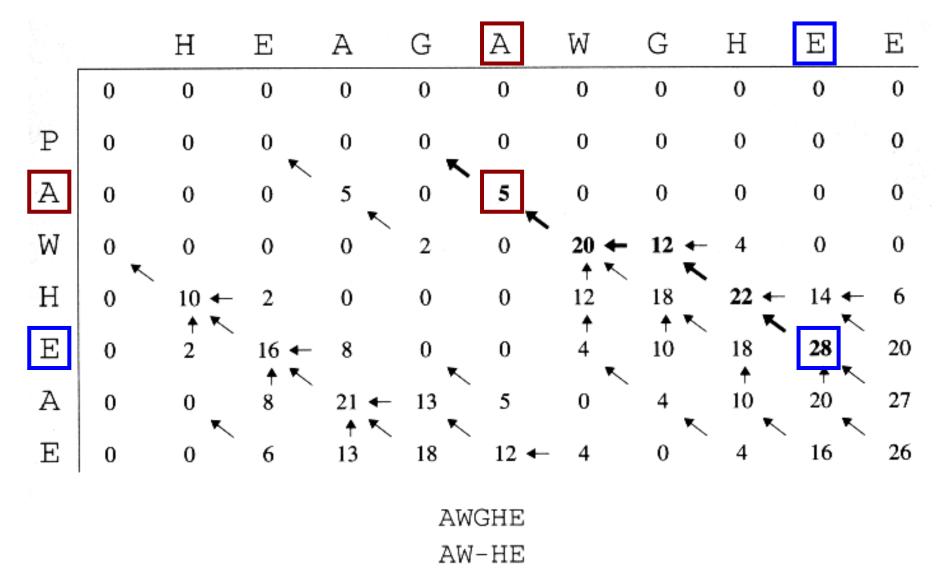
Recurrence relation:
$$F(i,j) = \max \begin{cases} 0\\F(i-1,j-1) + s(x_i,y_j)\\F(i-1,j) - d\\F(i,j-1) - d \end{cases}$$

Finally, find indices i^* and j^* after which the similarity only decreases. Stop the alignment there.

$$F(i^*, j^*) = \max_{i,j} F(i, j)$$

Traceback...

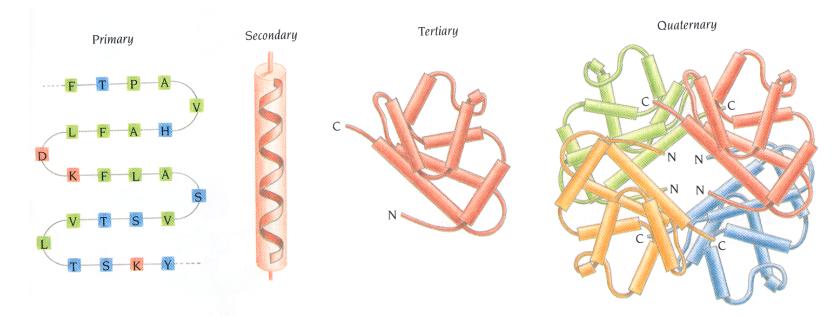
...starts at highest value until a cell with 0 is reached.



Adapted from Durbin et al., Cambridge University Press. https://doi.org/10.1017/CBO9780511790492.004

Local vs. Global Alignment: Biological Considerations

- Many proteins have **multiple domains**, or modules.
- Some domains are present (with high similarity) in many other proteins
- Local alignment can detect similar regions in otherwise dissimilar proteins.



Durbin et al., Cambridge University Press. https://doi.org/10.1017/CBO9780511790492.004

Other gap models

• So far: linear gap model. Not ideal for biological sequences, since it penalizes additional gap steps as much as the first. But in reality: When gaps do occur, they are often longer than one character.

HBA_HUMAN GSAQVKGHGKKVADALTNAVAHV---D--DMPNALSALSDLHAHKL ++ ++++H+ KV + +A ++ +L+ L+++H+ K LGB2_LUPLU NNPELQAHAGKVFKLVYEAAIQLQVTGVVVTDATLKNLGSVHVSKG

Durbin et al., Cambridge University Press. https://doi.org/10.1017/CBO9780511790492.004

• For a general gap cost function $\gamma(g)$, we can still use the standard dynamic programming recursion with slight modifications:

$$F(i,j) = \max \begin{cases} F(i-1, j-1) + s(x_i, y_j) \\ F(k, j) + \gamma(i-k), & k = 0, \dots, i-1, \\ F(i, k) + \gamma(j-k), & k = 0, \dots, j-1. \end{cases}$$

• **Problem:** requires $O(n^3)$ operations to align two sequences of length n, rather than $O(n^2)$. Why? \rightsquigarrow exercises...

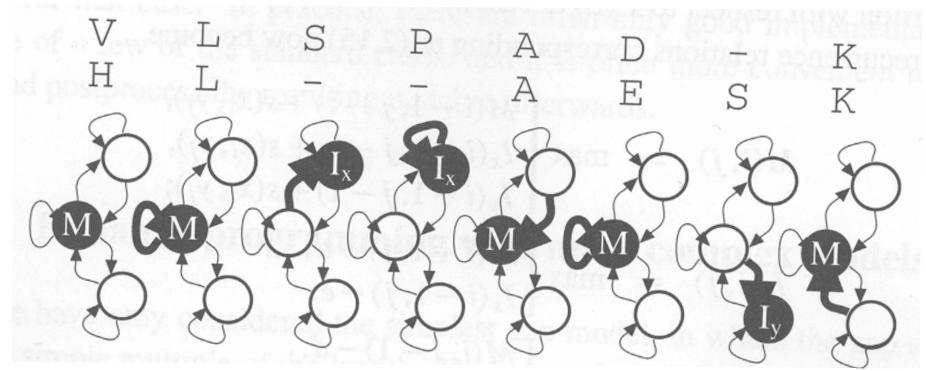
Alignment with affine gap costs

For affine gap costs, $\gamma(g) = -d - (g - 1)e$, there exists a **solution**: Modify recurrence by introducing another two "states". Denote by

- M(i, j) the best score given that x_i is aligned to y_j ,
- $I_x(i,j)$ the best score given that x_i is aligned to a gap,
- $I_y(i, j)$ the best score given that y_j is aligned to a gap.

$$M(i,j) = \max \begin{cases} M(i-1,j-1) + s(x_i, y_j) \\ I_x(i-1,j-1) + s(x_i, y_j) \\ I_y(i-1,j-1) + s(x_i, y_j) \end{cases} \overset{(x_i, y_j)}{=} \max \begin{cases} M(i-1,j) - d \\ I_x(i-1,j) - e \\ I_y(i,j) = \max \begin{cases} M(i,j-1) - d \\ I_y(i,j-1) - e \\ I_y(i,j-1) - e \end{cases} \overset{(x_i, y_j)}{=} \max \begin{cases} M(i,j-1) - d \\ I_y(i,j-1) - e \\ I_y(i,j-1) - e \end{cases}$$

Example FSA alignment



Durbin et al., Cambridge University Press. https://doi.org/10.1017/CBO9780511790492.004

FSA alignment corresponds to path through states.

Probabilistic version ~> Hidden Markov models (next chapter)