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Chapter III: Pairwise Sequence Alignment Part 2

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What will you learn?

To define homology as well as orthologs and paralogs

To explain how PAM (accepted point mutation) matrices are derived

To contrast the utility of PAM and BLOSUM scoring matrices.

To define dynamic programming and explain how global and local pairwise alignments are performed

To perform pairwise alignment of protein or DNA sequences at the NCBI website

Outline

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Introduction
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Protein alignment: often more informative than DNA alignment

Definitions: homology, similarity, identity

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Pairwise alignment, homology, and evolution of life

Scoring matrices

Dayhoff model: 7 steps

Pairwise alignment and limits of detection: the "twilight zone"

Alignment algorithms: global and local

Global sequence alignment: algorithm of Needleman and

Wunsch

Local sequence alignment: Smith and Waterman algorithm

Rapid, heuristic versions of Smith–Waterman: FASTA and BLAST

Basic Local Alignment Search Tool (BLAST)

Pairwise alignment with dotplots

The statistical significance of pairwise alignments

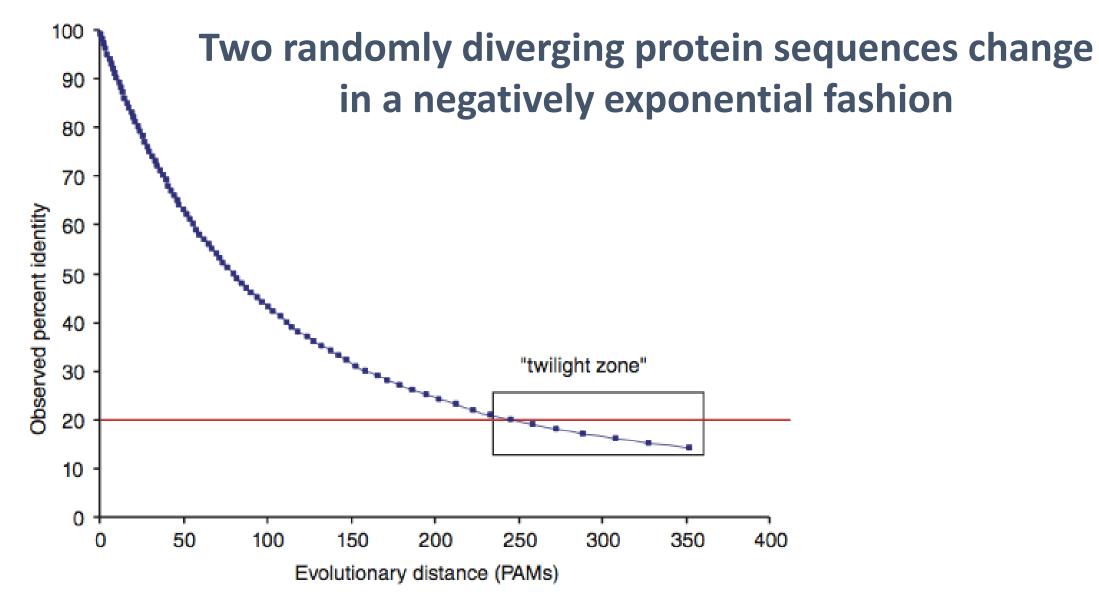
Statistical significance of global alignments

Percent identity and relative entropy

Perspective

PAM matrices: Point-accepted mutations

- > PAM matrices are based on global alignments of closely related proteins.
- The PAM1 is the matrix calculated from comparisons of sequences with no more than 1% divergence. At an evolutionary interval of PAM1, one change has occurred over a length of 100 amino acids.
- ➤ Other PAM matrices are extrapolated from PAM1. For PAM250, 250 changes have occurred for two proteins over a length of 100 amino acids.
- ➤ All the PAM data come from closely related proteins (>85% amino acid identity).



- At PAM1, two proteins are 99% identical At PAM10.7, there are 10 differences per 100 residues At PAM80, there are 50 differences per 100 residues At PAM250, there are 80 differences per 100 residues

Correspondence between observed differences (per 100 residues) and evolutionary distance (in PAMs)

Observed differences in 100 residues	Evolutionary distance in PAMs	
1	1.0	
5	5.1	
10	10.7	
15	16.6	
20	23.1	
25	30.2	
30	38.0	
35	47	
40	56	
45	67	7
50	80	PAM 80
55	94	
60	112	
65	133	
70	159	
75	195	1
80	246	PAM 250

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Two kinds of sequence alignment: global and local

- ➤ We will first consider the global alignment algorithm of Needleman and Wunsch (1970).
- ➤ We will then explore the local alignment algorithm of Smith and Waterman (1981).
- Finally, we will consider BLAST, a heuristic version of Smith-Waterman.

Global alignment with the algorithm of Needleman and Wunsch (1970)

- Two sequences can be compared in a matrix along x- and y-axes.
- If they are identical, a path along a diagonal can be drawn.
- Find the optimal subpaths and add them up to achieve the best score. This involves:
 - -adding gaps when needed;
 - -allowing for conservative substitutions;
 - -choosing a scoring system (simple or complicated).
- N-W is guaranteed to find optimal alignment(s).

Three steps to global alignment with the Needleman-Wunsch algorithm

- [I] set up a matrix
- [2] score the matrix
- [3] identify the optimal alignment(s)

Four possible outcomes in aligning two sequences

sequence 1 (length m)

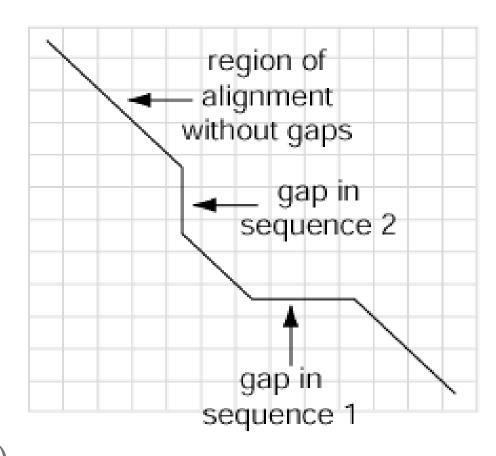
sequence 2 (length n)

[1] identity (stay along a diagonal)

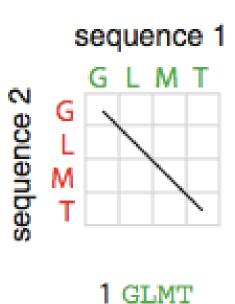
[2] mismatch (stay along a diagonal)

[3] gap in one sequence (move vertically!)

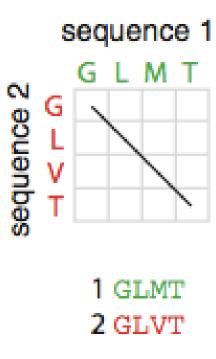
[4] gap in the other sequence (move horizontally!)

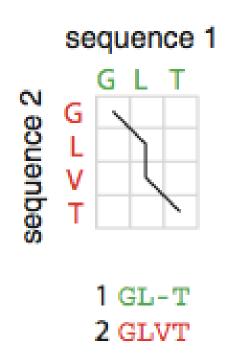


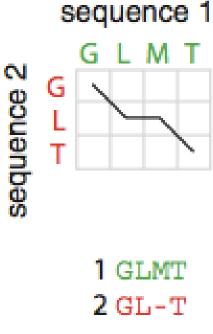
Four possible outcomes in aligning two sequences



2 GLMT

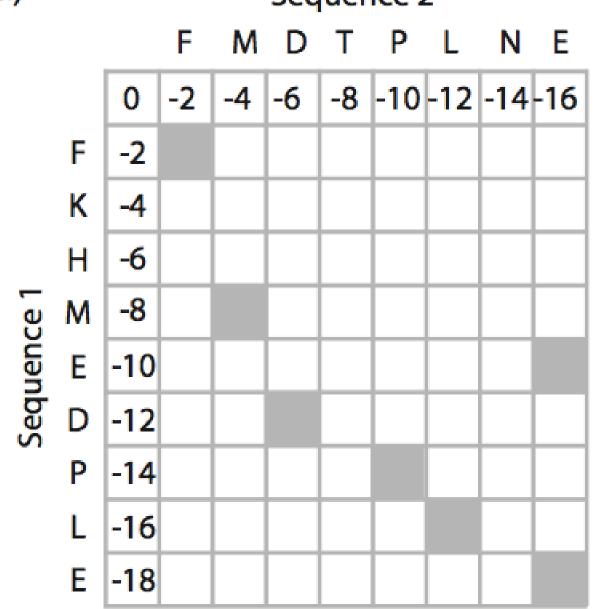






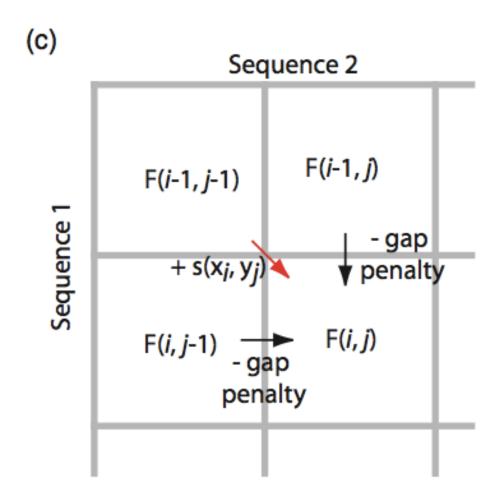
Global pairwise alignment using Needleman-Wunsch (a) Sequence 2

Identify positions of identity (shaded gray).

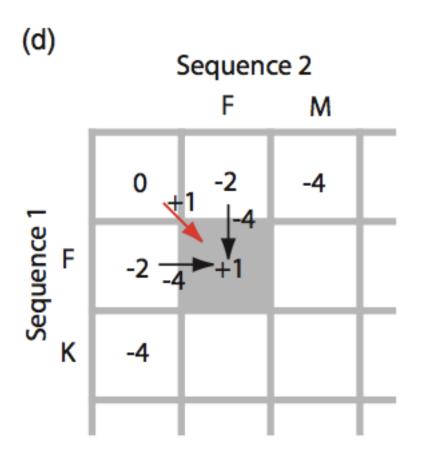


Score = Max
$$\begin{cases} F(i-1, j-1) + s(x_i, y_i) \\ F(i-1, j) - gap penalty \\ F(i, j-1) - gap penalty \end{cases}$$

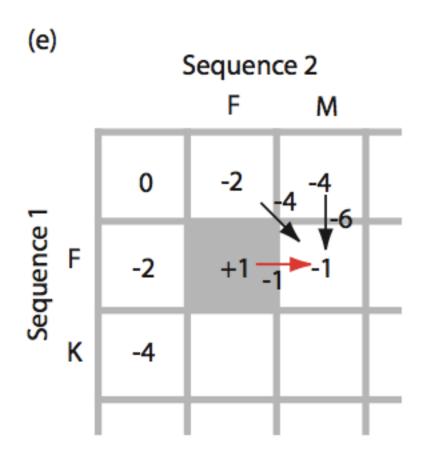
Define an overall score that maximizes cumulative scores at each position of the pairwise alignment, allowing for substitutions and gaps in either sequence.



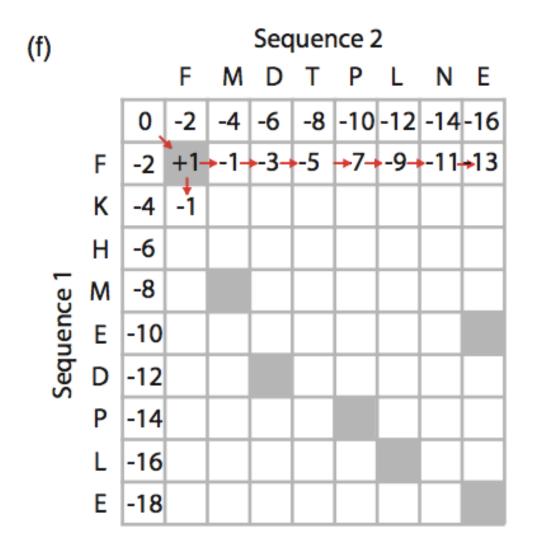
To decide how to align sequences 1 and 2 in the box at lower right, decide what the scores are beginning at upper left (not requiring a gap), or beginning from the left or top (each requiring a gap penalty).



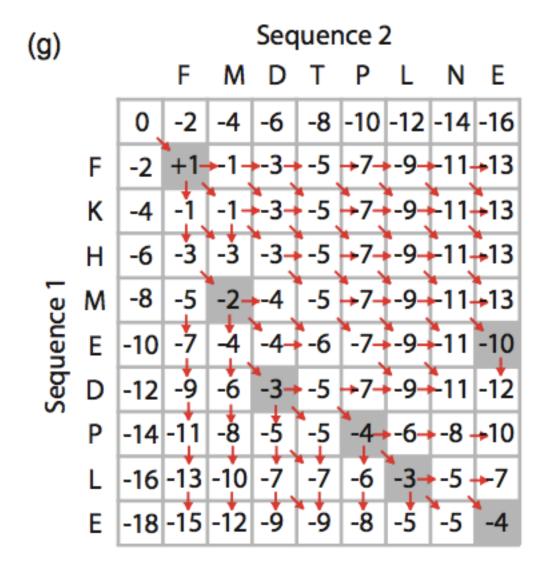
Here the best score involves +1 (proceed from upper left to gray, lower right square). If we instead select an alignment involving a gap the score would be worse (-4).



Proceed to calculate the optimal score for the next position.

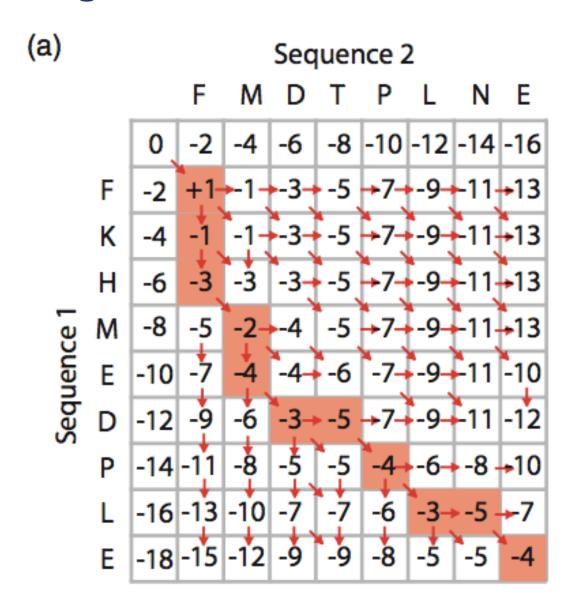


Continue filling in the matrix.

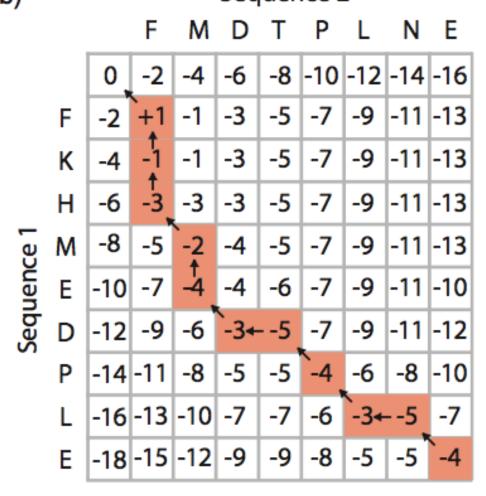


Complete filling in the matrix.

Highlighted cells indicate the optimal path (best scores), indicating how the two sequences should be aligned.

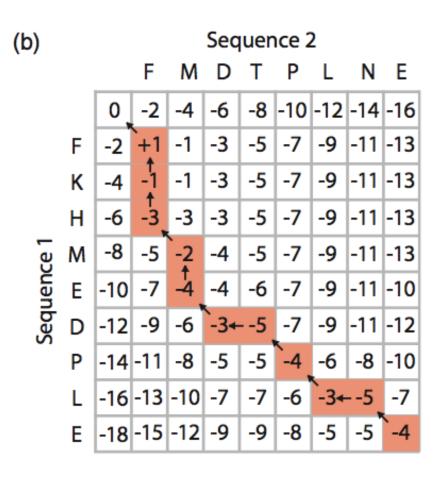


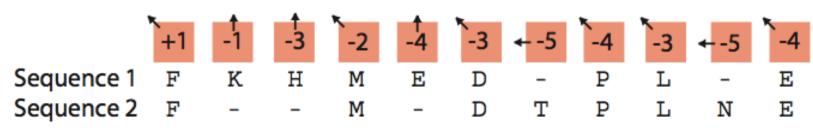
Global pairwise alignment using Needleman-Wunsch (b) Sequence 2



B&FG 3e Fig. 3-22 Page 100 Equivalent representation, showing the traceback procedure: begin at the lower right cell and proceed back to the start.

Equivalent representation, showing the traceback procedure: begin at the lower right cell and proceed back to the start.





Needleman-Wunsch: dynamic programming

N-W is guaranteed to find optimal alignments, although the algorithm does not search all possible alignments.

It is an example of a dynamic programming algorithm: an optimal path (alignment) is identified by incrementally extending optimal subpaths. Thus, a series of decisions is made at each step of the alignment to find the pair of residues with the best score.

Global alignment versus local alignment

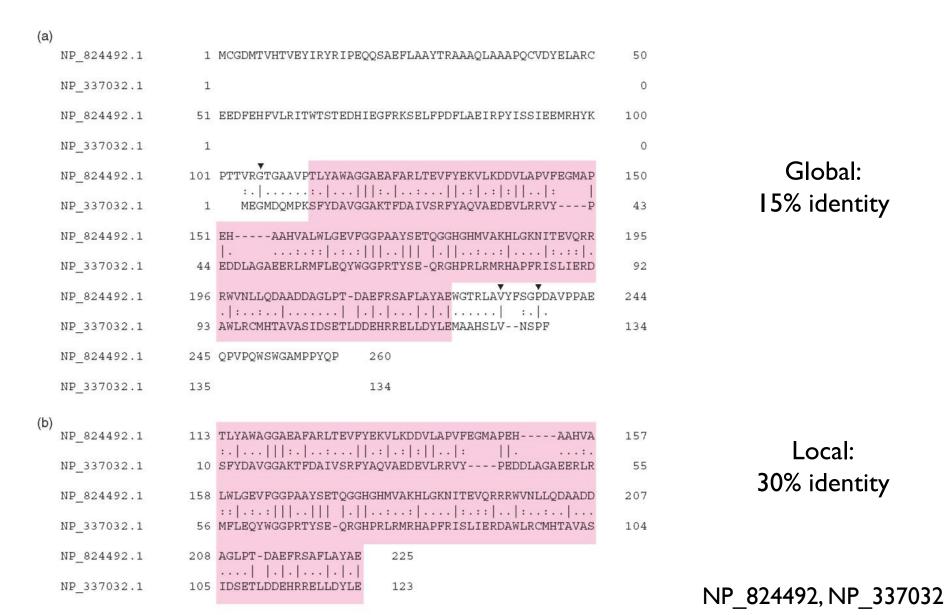
Global alignment (Needleman-Wunsch) extends from one end of each sequence to the other.

Local alignment finds optimally matching regions within two sequences ("subsequences").

Local alignment is almost always used for database searches such as **BLAST**. It is useful to find domains (or limited regions of homology) within sequences.

Smith and Waterman (1981) solved the problem of performing optimal local sequence alignment. Other methods (BLAST, FASTA) are faster but less thorough.

Global alignment (top) includes matches ignored by local alignment (bottom)



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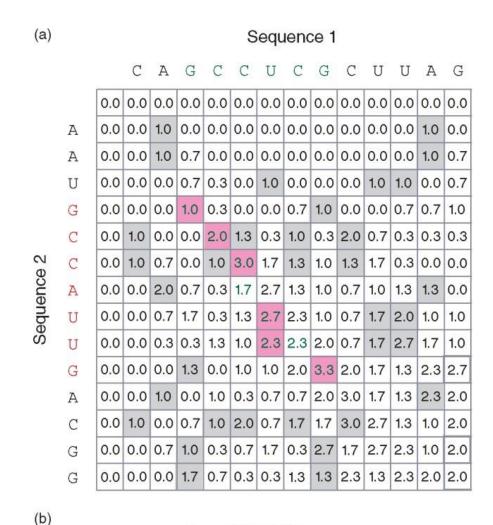
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Percent identity and relative entropy

Perspective

How the Smith-Waterman algorithm works



GCC-UCG

GCCAUUG

CA-GCC-UCGCUUAG

AAUGCCAUUGACG-G

sequence 1

sequence 2

sequence 1

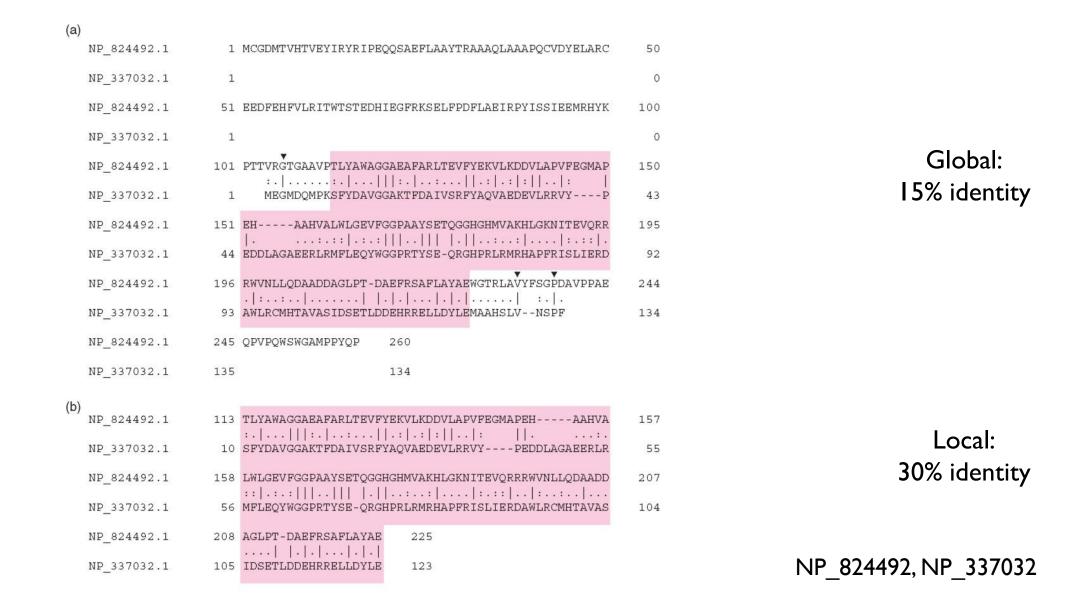
sequence 2

(c)

positions of nucleotide identity are shaded gray

They scoring system here is +1 for a match, minus one-third for a mismatch, and a gap penalty of the difference between a match and a mismatch (-1.3 for a gap of length one). The matrix is scored based on finding the maximum of four possible non-negative values. The highest value in the matrix (3.3) corresponds to the beginning of the optimal local alignment, and the aligned residues (green font) extend up and to the left until a value of zero is reached. (b) The local alignment derived from this matrix is shown. Note that this alignment includes identities, a mismatch, and a gap. (c) A global alignment of the two sequences is shown for comparison to the local alignment. Note that it encompasses the entirety of both sequences.

Global alignment (top) includes matches ignored by local alignment (bottom)



Where to use the Smith-Waterman algorithm

- [1] Galaxy offers "needle" and "water" EMBOSS programs.
- [2] EBI offers needle and water. http://www.ebi.ac.uk/Tools/psa/
- [3] Try using **SSEARCH** to perform a rigorous Smith-Waterman local alignment: http://fasta.bioch.virginia.edu/
- [4] Next-generation sequence aligners incorporate Smith-Waterman in some specialized steps.

Rapid, heuristic versions of Smith-Waterman: FASTA and BLAST

- <u>Smith-Waterman is very rigorous</u>, and it is guaranteed to find an optimal alignment.
- But Smith-Waterman <u>is slow</u>. It requires computer space and time proportional to the product of the two sequences being aligned (or the product of a query against an entire database).
- Gotoh (1982) and Myers and Miller (1988) improved the algorithms so both global and local alignment require less time and space.
- FASTA and BLAST provide rapid alternatives to S-W.

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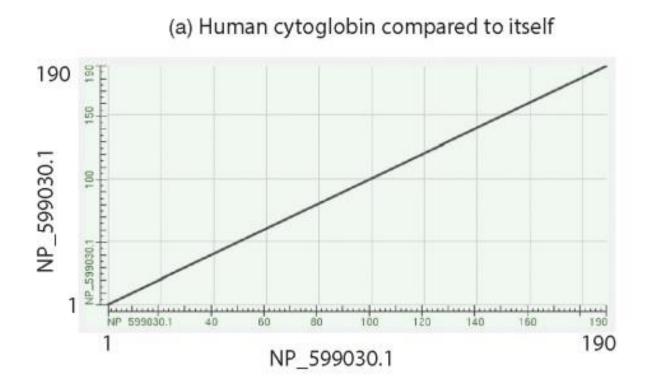
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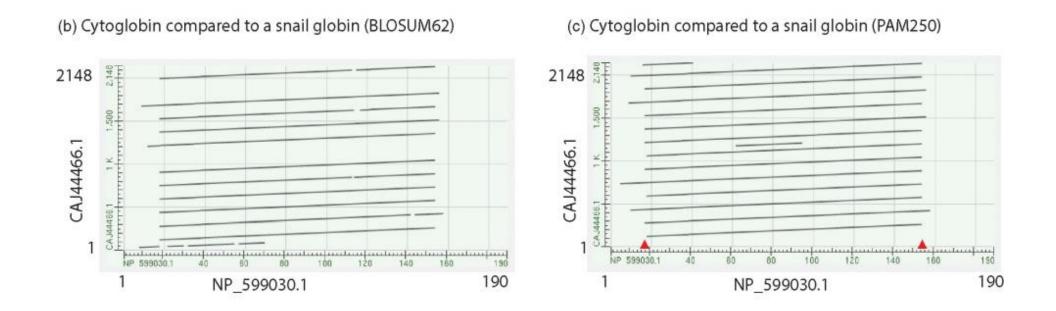
Perspective

Pairwise alignment with dotplots



A human globin searched against itself produces a unit diagonal on a dot plot (NCBI BLASTP, aligning 2 sequences).

Pairwise alignment with dotplots



Search human cytoglobin against a large snail globin (having many globin repeats). More repeats are observed using PAM250 than BLOSUM62.

To "read" this plot note that cytoglobin (x-axis) matches the snail globin (y-axis) at about a dozen locations across the snail protein. Red arrows indicate that the first few and last few amino acids of cytoglobin do not participate in this repeat structure.

Pairwise alignment with dotplots

BLASTP output includes the various sequence alignments. One is shown here: human cytoglobin (residues 18-154) aligns to the snail globin (at residues 1529-1669). The expect value is convincing (4e-13), and this is one of a dozen sequence alignments.

Conclusion: the dotplot is an excellent way to visualize complex repeats.

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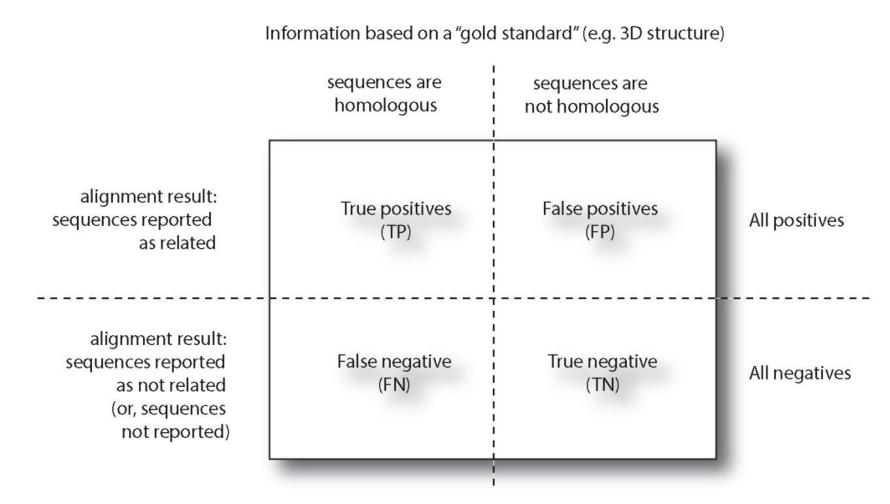
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Statistical significance of pairwise alignments



Sensitivity = TP / (TP + FN) Specificity = TN / (TN + FP)

Statistical significance of pairwise alignments

The statistical significance of global alignments is not well described. We can apply a z-score.

$$Z = \frac{x - \mu}{s}$$

For local alignment the statistical significance is thoroughly understood. See Chapter 4 (BLAST).

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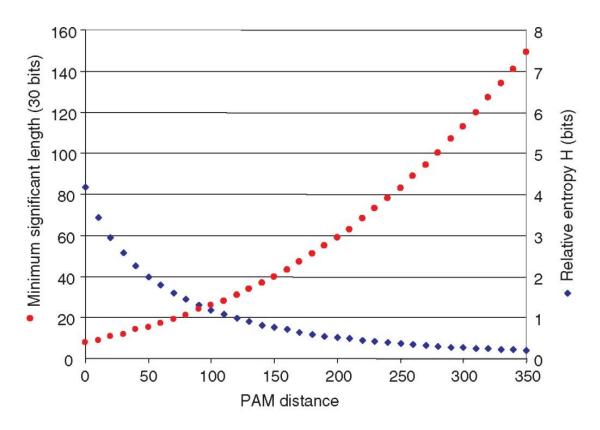
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Relative entropy (H) as a function of PAM distance



For **PAM** matrices with low values (e.g. **PAM10**) the relative entropy (in bits) is high, and the minimum alignment length needed to detect a significant pairwise alignment is short. Relative entropy relates to the information content.

For **PAM250** and similar matrices the relative entropy is low. It is necessary to have a longer region of amino acids aligned (e.g. 80 residues) to detect significant pairwise relatedness.

Perspective: Pairwise alignment is a fundamental problem in bioinformatics.