# Sequence alignment

 Alignment specifies which positions in two sequences match

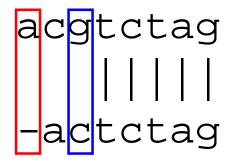
acgtctag	acgtctag	acgtctag
actctag-	-actctag	ac-tctag

2 matches 5 mismatches 1 not aligned

5 matches 2 mismatches 1 not aligned 7 matches0 mismatches1 not aligned

# Mutations: Insertions, deletions and substitutions

Indel: insertion or deletion of a base with respect to the ancestor sequence



Mismatch: substitution (point mutation) of a single base

- Insertions and/or deletions are called *indels* 
  - We can't tell whether the ancestor sequence had a base or not at indel position

#### Problems

- What sorts of alignments should be considered?
- How to score alignments?
- How to find optimal or good scoring alignments?
- How to evaluate the statistical significance of scores?

In this course, we discuss each of these problems briefly.

Course *Biological sequence analysis* tackles all four indepth.

#### Sequence Alignment (chapter 6)

- The biological problem
- Global alignment
- Local alignment
- Multiple alignment

# Global alignment

- Problem: find optimal scoring alignment between two sequences (Needleman & Wunsch 1970)
- Every position in both sequences is included in the alignment
- We give score for each position in alignment

<ul> <li>Identity (match)</li> </ul>	+1	WHAT
<ul> <li>Substitution (mismatch)</li> </ul>	-µ	
– Indel	-δ	WH-Y

Total score: sum of position scores

 $S(WHAT/WH-Y) = 1 + 1 - \delta - \mu$ 

# Dynamic programming

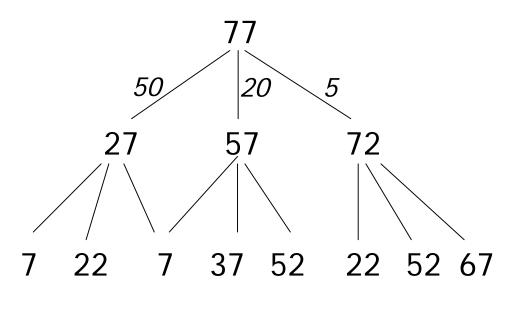
- How to find the optimal alignment?
- We use previous solutions for optimal alignments of smaller subsequences
- This general approach is known as dynamic programming

Introduction to dynamic programming: the money change problem

- Suppose you buy a pen for 4.23€ and pay for with a 5€ note
- You get 77 cents in change what coins is the cashier going to give you if he or she tries to minimise the number of coins?
- The usual algorithm: start with largest coin (denominator), proceed to smaller coins until no change is left:
  - 50, 20, 5 and 2 cents
- This greedy algorithm is *incorrect*, in the sense that it does not always give you the correct answer

#### The money change problem

- How else to compute the change?
- We could consider all possible ways to reduce the amount of change
- Suppose we have 77 cents change, and the following coins: 50, 20, 5 cents
- We can compute the change with recursion
- Figure shows the recursion tree for the example

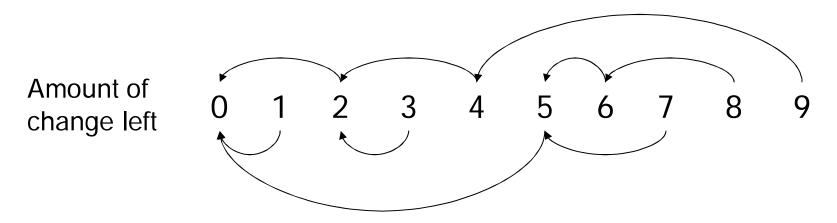


- Many values are computed more than once!
- This leads to a correct but very inefficient algorithm

#### The money change problem

- We can speed the computation up by solving the change problem for all  $i \le n$ 
  - Example: solve the problem for 9 cents with available coins being 1, 2 and 5 cents
- Solve the problem in steps, first for 1 cent, then 2 cents, and so on
- In each step, utilise the solutions from the previous steps

#### The money change problem



- Algorithm runs in time proportional to Md, where M is the amount of change and d is the number of coin types
- The same technique of storing solutions of subproblems can be utilised in aligning sequences

# Representing alignments and scores

Alignments can be represented in the following tabular form.

Each alignment corresponds to a path through the table.

#### WHAT

-W X H X X Y X X

W

Η

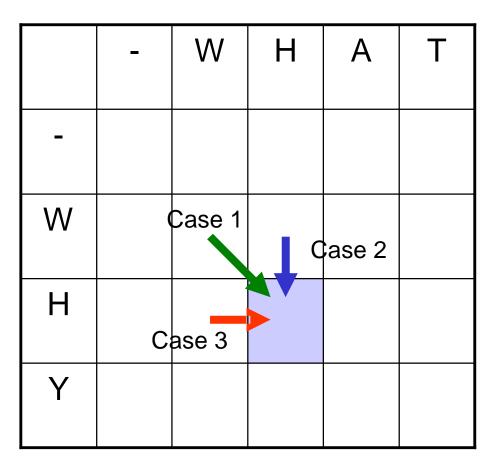
Α

#### WH-Y

#### Representing alignments and scores

W Н Α Т WHAT 0 WH-Y W 1 Η 2 2-δ **Global alignment** Y 2-δ-μ score  $S_{3,4} = 2 - \delta - \mu$ 

### Filling the alignment matrix



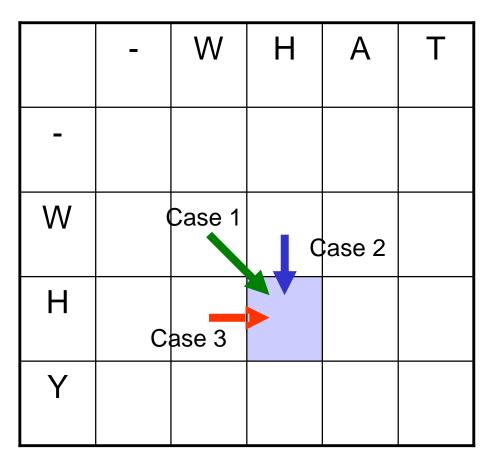
Consider the alignment process at shaded square.

Case 1. Align H against H (match or substitution).

Case 2. Align H in WHY against – (indel) in WHAT.

Case 3. Align H in WHAT against – (indel) in WHY.

# Filling the alignment matrix (2)



Scoring the alternatives. Case 1.  $S_{2,2} = S_{1,1} + s(2, 2)$ Case 2.  $S_{2,2} = S_{1,2} - \delta$ Case 3.  $S_{2,2} = S_{2,1} - \delta$ s(i, j) = 1 for matching positions,  $s(i, j) = -\mu$  for substitutions. Choose the case (path) that yields the maximum score.

Keep track of path choices.

# Global alignment: formal development

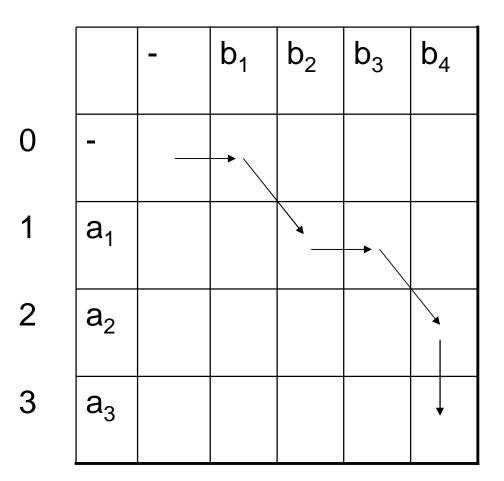
 $A = a_1 a_2 a_3 \dots a_n,$   $B = b_1 b_2 b_3 \dots b_m$  $b_1 \quad b_2 \quad b_3 \quad b_4$ 

 $b_1 \quad b_2 \quad b_3 \quad b_4 \quad -$ -  $a_1 \quad - \quad a_2 \quad a_3$ 

Any alignment can be written as a unique path through the matrix

Score for aligning A and B up to positions i and j:

$$S_{i,j} = S(a_1a_2a_3...a_i, b_1b_2b_3...b_j)$$



# Scoring partial alignments

- Alignment of  $A = a_1 a_2 a_3 \dots a_n$  with  $B = b_1 b_2 b_3 \dots b_m$  can end in three ways
  - Case 1:  $(a_1a_2...a_{i-1}) a_i$   $(b_1b_2...b_{j-1}) b_j$ - Case 2:  $(a_1a_2...a_{i-1}) a_i$   $(b_1b_2...b_j) -$ - Case 3:  $(a_1a_2...a_i) (b_1b_2...b_{j-1}) b_j$

#### Scoring alignments

1 if a = b

Scores for each case:

I

$$- \text{ Case 1: } (a_1 a_2 \dots a_{i-1}) a_i \qquad s(a_i, b_j) = \begin{cases} +1 \text{ If } a_i = b_j \\ -\mu \text{ otherwise} \end{cases}$$

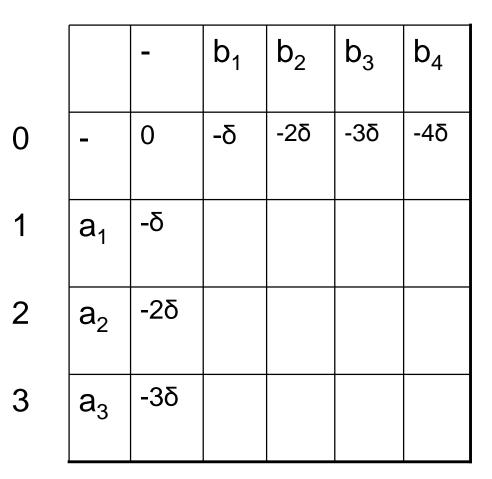
$$- \text{ Case 2: } (a_1 a_2 \dots a_{i-1}) a_i \qquad (b_1 b_2 \dots b_j) - \\ - \text{ Case 3: } (a_1 a_2 \dots a_i) - \qquad (b_1 b_2 \dots b_{j-1}) b_j \end{cases}$$

# Scoring alignments (2)

 First row and first column correspond to initial alignment against indels:

$$\begin{split} S(i, 0) &= -i \ \delta \\ S(0, j) &= -j \ \delta \end{split}$$

 Optimal global alignment score S(A, B) = S<sub>n,m</sub>



### Algorithm for global alignment

```
Input sequences A, B, n = |A|, m = |B|
Set S<sub>i,0</sub> := -\deltai for all i
Set S<sub>0,j</sub> := -\deltaj for all j
for i := 1 to n
for j := 1 to m
S<sub>i,j</sub> := max{S<sub>i-1,j</sub> - \delta, S<sub>i-1,j-1</sub> + s(a<sub>i</sub>,b<sub>j</sub>), S<sub>i,j</sub>-1 - \delta}
end
end
```

#### Algorithm takes O(nm) time and space.

#### Global alignment: example

μ = 1 δ = 2

	-	Т	G	G	Т	G
-	0	-2	-4	-6	-8	-10
А	-2					
Т	-4					
С	-6					
G	-8					
Т	-10					?

#### Global alignment: example (2)

Т

-2

-

()

G

-4

G

-6

Т

-8

G

-10

-9

-6

-5

-4

-2

 $\mu = 1$  $\delta = 2$ 

ATCGT-

-2 -3 Α -1 -5 -7 × -2 Т -4 -1 -4 -4 С -6 -3 -2 -3 -5  $\mathbf{\lambda}$ G -2 -8 -5 -3 -1 X Т -10 -7 -3 -4 0

-TGGTG

#### Sequence Alignment (chapter 6)

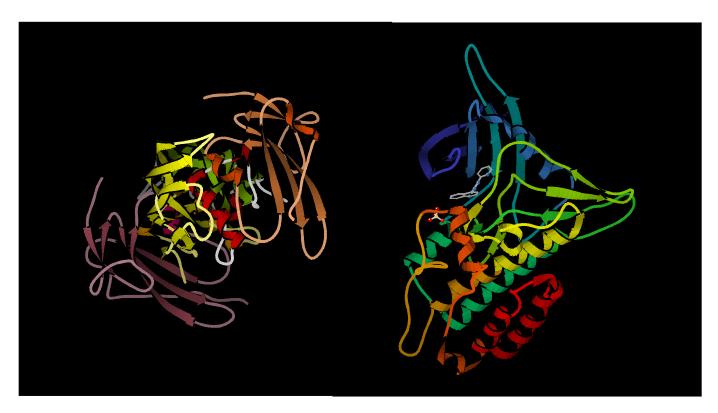
- The biological problem
- Global alignment
- Local alignment
- Multiple alignment

# Local alignment: rationale

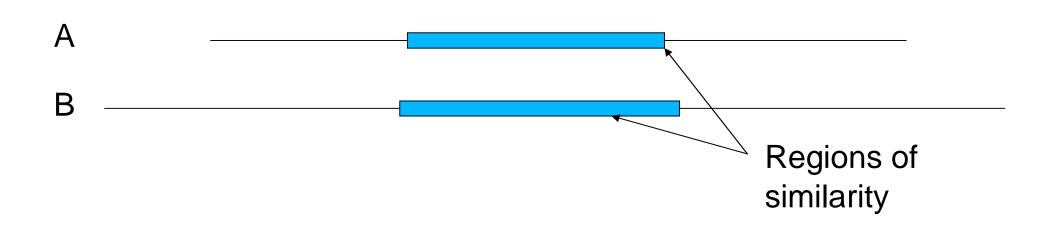
- Otherwise dissimilar proteins may have local regions of similarity
  - -> Proteins may share a function

Human bone morphogenic protein receptor type II precursor (left) has a 300 aa region that resembles 291 aa region in TGF-β receptor (right).

The shared function here is protein kinase.



# Local alignment: rationale



- Global alignment would be inadequate
- Problem: find the highest scoring *local* alignment between two sequences
- Previous algorithm with minor modifications solves this problem (Smith & Waterman 1981)

### From global to local alignment

- Modifications to the global alignment algorithm
  - Look for the highest-scoring path in the alignment matrix (not necessarily through the matrix), or in other words:
  - Allow preceding and trailing indels without penalty

#### Scoring local alignments

 $A = a_1 a_2 a_3 \dots a_n, B = b_1 b_2 b_3 \dots b_m$ 

Let I and J be intervals (substrings) of A and B, respectively:  $I \subset A$ ,  $J \subset B$ 

Best local alignment score:

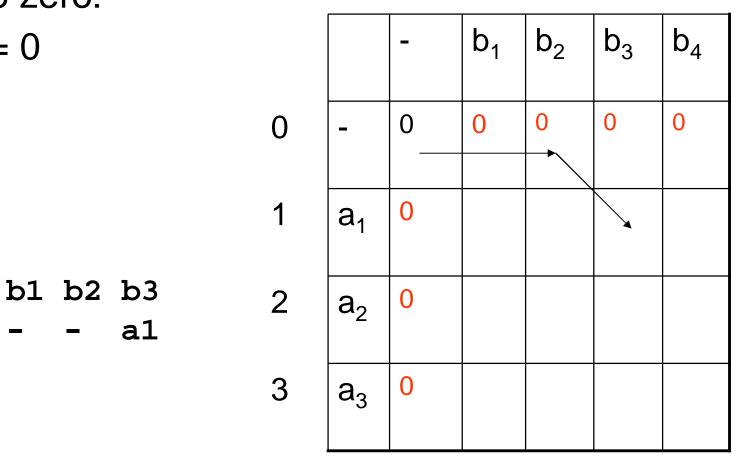
 $M(A,B) = \max\{S(I,J) : I \subset A, J \subset B\}$ 

where S(I, J) is the score for substrings I and J.

# Allowing preceding and trailing indels

 First row and column initialised to zero:

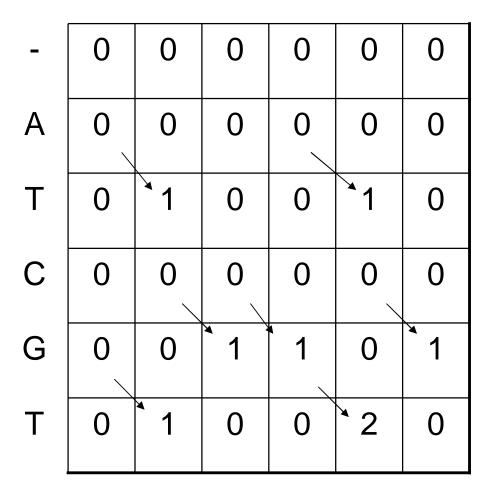
$$\mathsf{M}_{i,0} = \mathsf{M}_{0,j} = \mathsf{O}$$



a1

#### Recursion for local alignment

• 
$$M_{i,j} = \max \{$$
  
 $M_{i-1,j-1} + s(a_i, b_i),$   
 $M_{i-1,j} - \delta,$   
 $M_{i,j-1} - \delta,$   
 $0$   
}



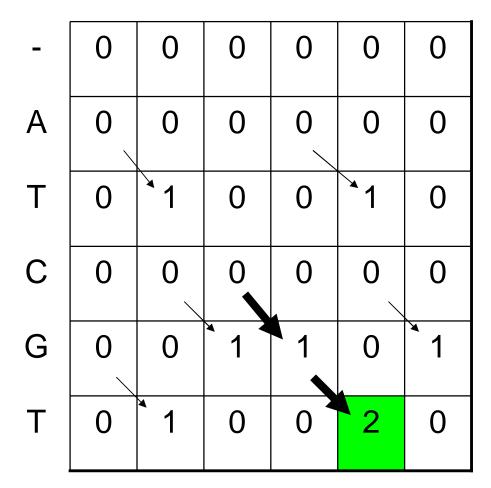
# Finding best local alignment

 Optimal score is the highest value in the matrix

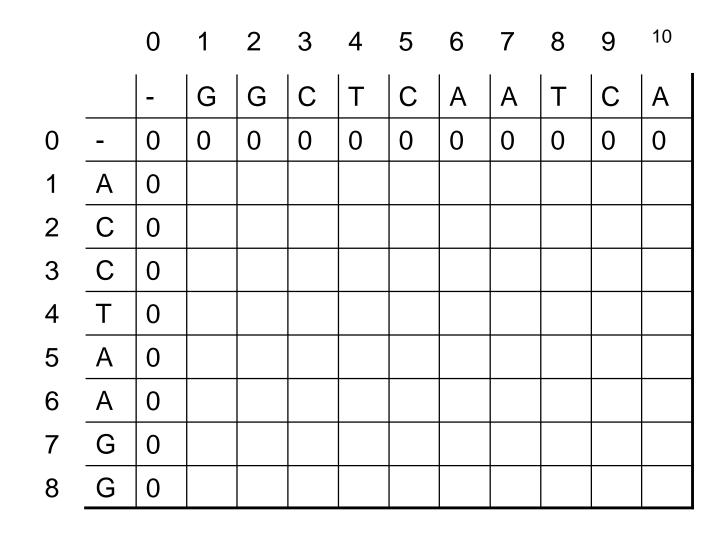
 $M(A,B) = \max\{S(I,J) : I \subset A, J \subset B\}$ 

 $= \max_{i,j} M_{i,j}$ 

 Best local alignment can be found by backtracking from the highest value in M



#### Local alignment: example



#### Local alignment: example

			0	1	2	3	4	5	6	7	8	9	10
Scoring			-	G	G	C	T	C	A	A	T	C	A
Match: +2	0	-	0	0	0	0	0	0	0	0	0	0	0
Mismatch: -1	1	Α	0	0	0	0	0	0	2	2	0	0	2
	2	С	0	0	0 、	2	0	2	0	1	1	2	0
Indel: -2	3	С	0	0	0	2	1	2	1	0	0	3	1
	4	Т	0	0	0	0	4 -	2	1	0	2	1	2
	5	А	0	0	0	0	2	3	4	3	1	1	3
	6	А	0	0	0	0	0	1	5	<b>*</b> 6	4	2	3
СТ-АА	7	G	0	2	2	0	0	0	3	4	5	3	1
CTCAA	8	G	0	2	4	2	0	0	1	2	3	4	2

#### Non-uniform mismatch penalties

We used uniform penalty for mismatches:

 $s(A', C') = s(A', G') = \dots = s(G', T') = \mu$ 

- Transition mutations (A->G, G->A, C->T, T->C) are approximately twice as frequent than transversions (A->T, T->A, A->C, G->T)
  - use non-uniform mismatch
     penalties collected into a
     substitution matrix

	Α	С	G	Т
A	1	-1	-0.5	-1
С	-1	<b>~</b>	-1	-0.5
G	-0.5	-1	1	-1
Т	-1	-0.5	-1	1

# Gaps in alignment

Gap is a succession of indels in alignment

- Previous model scored a length k gap as  $w(k) = -k\delta$
- Replication processes may produce longer stretches of insertions or deletions
  - In coding regions, insertions or deletions of codons may preserve functionality

# Gap open and extension penalties (2)

We can design a score that allows the penalty opening gap to be larger than extending the gap:

$$w(k) = -\alpha - \beta(k-1)$$

- Gap open cost  $\alpha$ , Gap extension cost  $\beta$
- Our previous algorithm can be extended to use w(k)
   (not discussed on this course)

### Amino acid sequences

- We have discussed mainly dna sequences
- Amino acid sequences can be aligned as well
- However, the design of the substitution matrix is more involved because of the larger alphabet
- More on the topic in the course Biological sequence analysis

#### Sequence Alignment (chapter 6)

- The biological problem
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- Local alignment
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# Multiple alignment

- Consider a set of n sequences on the right
  - Orthologous sequences from different organisms
  - Paralogs from multiple duplications
- How can we study relationships between these sequences?

aggcgagctgcgagtgcta cgttagattgacgctgac ttccggctgcgac gacacggcgaacgga agtgtgcccgacgagcgaggac gcgggctgtgagcgcta aagcggcctgtgtgcccta atgctgctgccagtgta agtcgagccccgagtgc agtccgagtcc actcggtgc

# Optimal alignment of three sequences

- Alignment of  $A = a_1 a_2 \dots a_i$  and  $B = b_1 b_2 \dots b_j$  can end either in (-,  $b_j$ ),  $(a_i, b_j)$  or  $(a_i, -)$
- $2^2 1 = 3$  alternatives
- Alignment of A, B and C =  $c_1c_2...c_k$  can end in  $2^3 1$ ways:  $(a_i, -, -)$ ,  $(-, b_j, -)$ ,  $(-, -, c_k)$ ,  $(-, b_j, c_k)$ ,  $(a_i, -, c_k)$ ,  $(a_i, b_j, -)$  or  $(a_i, b_j, c_k)$
- Solve the recursion using three-dimensional dynamic programming matrix: O(n<sup>3</sup>) time and space
- Generalizes to n sequences but impractical with moderate number of sequences

# Multiple alignment in practice

- In practice, real-world multiple alignment problems are usually solved with heuristics
- Progressive multiple alignment
  - Choose two sequences and align them
  - Choose third sequence w.r.t. two previous sequences and align the third against them
  - Repeat until all sequences have been aligned
  - Different options how to choose sequences and score alignments

# Multiple alignment in practice

- Profile-based progressive multiple alignment:
   CLUSTALW
  - Construct a distance matrix of all pairs of sequences using dynamic programming
  - Progressively align pairs in order of decreasing similarity
  - CLUSTALW uses various heuristics to contribute to accuracy

### Additional material

- R. Durbin, S. Eddy, A. Krogh, G. Mitchison: Biological sequence analysis
- N. C. Jones, P. A. Pevzner: An introduction to bioinformatics algorithms
- Course Biological sequence analysis in Spring 2008

#### Demonstration of the EBI web site

 European Bioinformatics Institute (EBI) offers many biological databases and bioinformatics tools at http://www.ebi.ac.uk/