Sequence alignments and scoring matrices

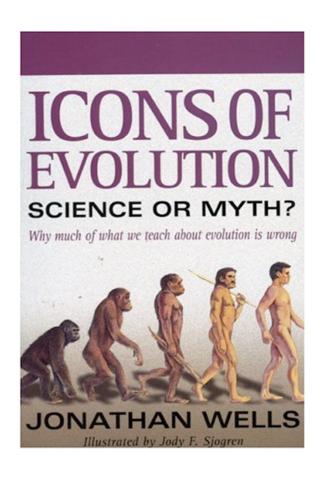
Arne Elofsson

To read: http://perso.fundp.ac.be/~lambertc/DEA-bioinfo/CLambert_curr_gen_2003.pdf

To read: Wikipidea about Sequence Alignment

Why alignments?

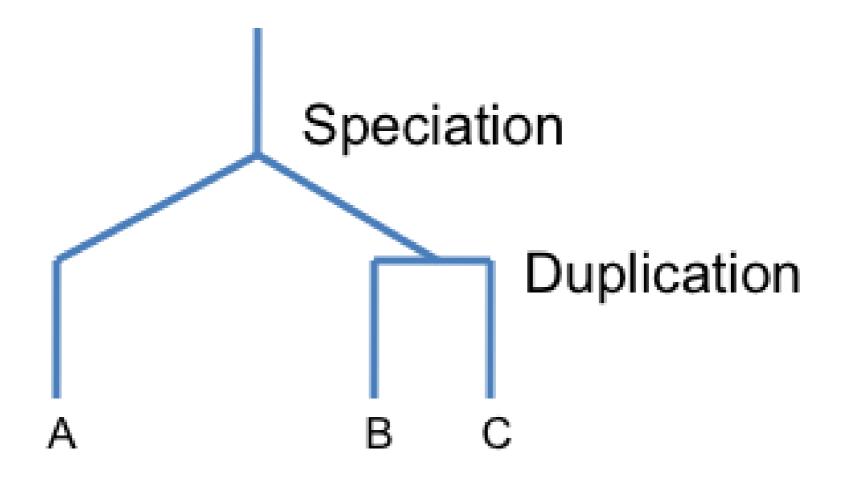
- Detect homology
- Study evolution
- Predict functions
- Model 3D-structure



Sequence similarity

- Homologs have a common ancestor
- Gene duplication or speciation
- High sequence similarity indicates homology
- Homologs have similar 3D-structure

Homology



Convergent evolution

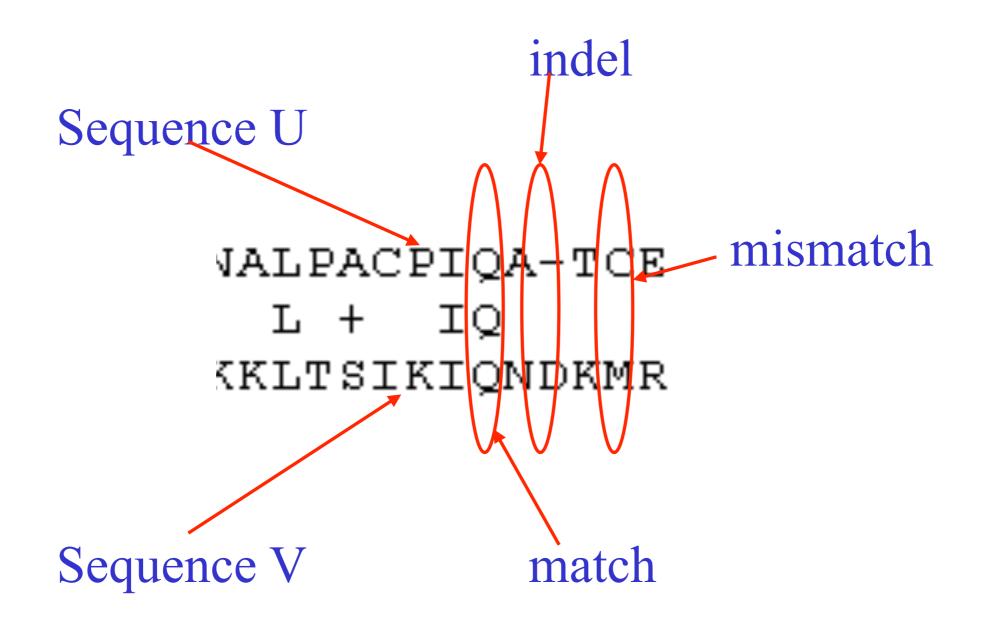




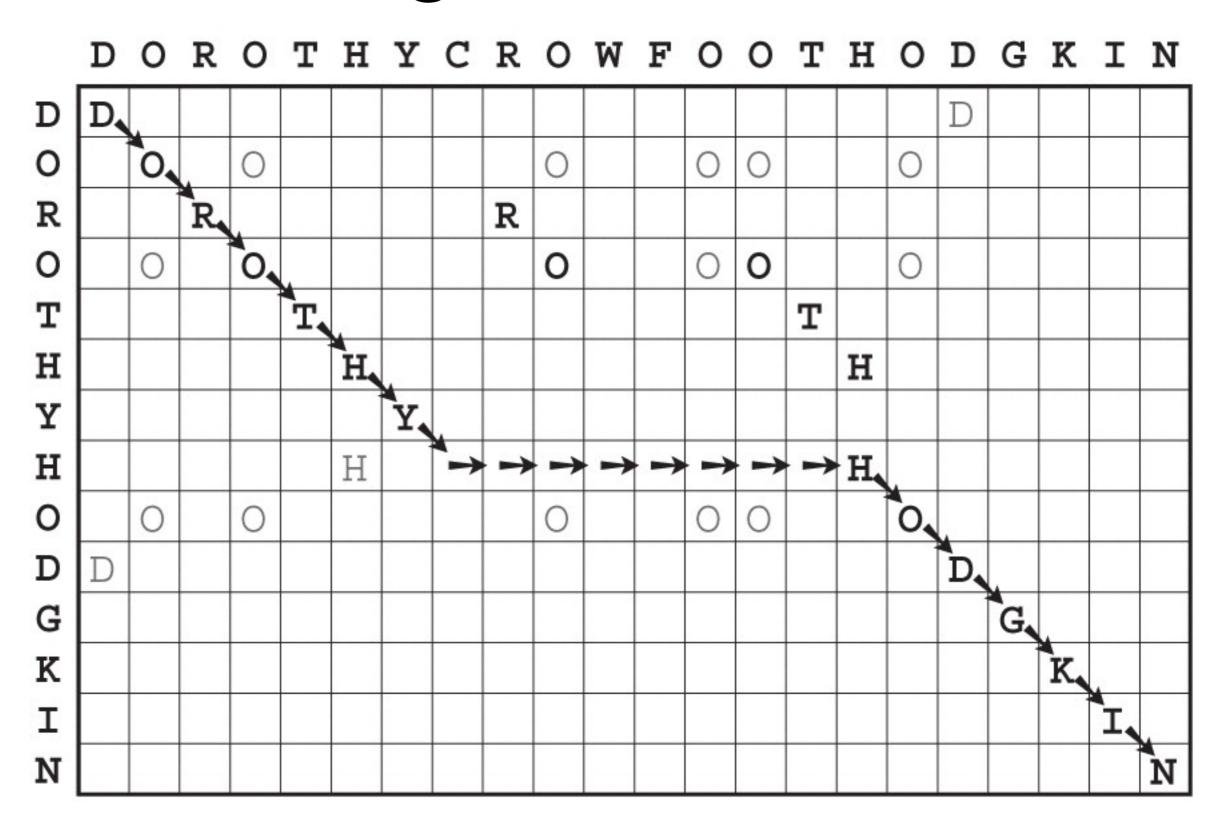
What is an alignment

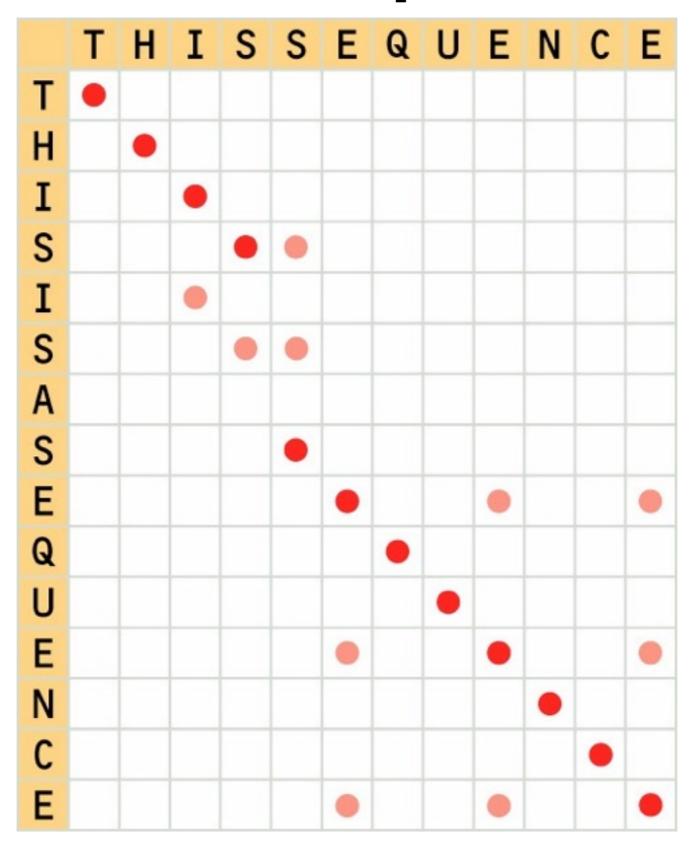
```
THISSEQUENCE
                 10/12 Identical
THATSEQUENCE
THATSEQUENCE
   | | 4/12 Identical
THISISASEQUENCE
THISISA-SEQUENCE
      | | | | | | | | | 11/12 Identical
TH----ATSEQUENCE
```

What can an alignment say?

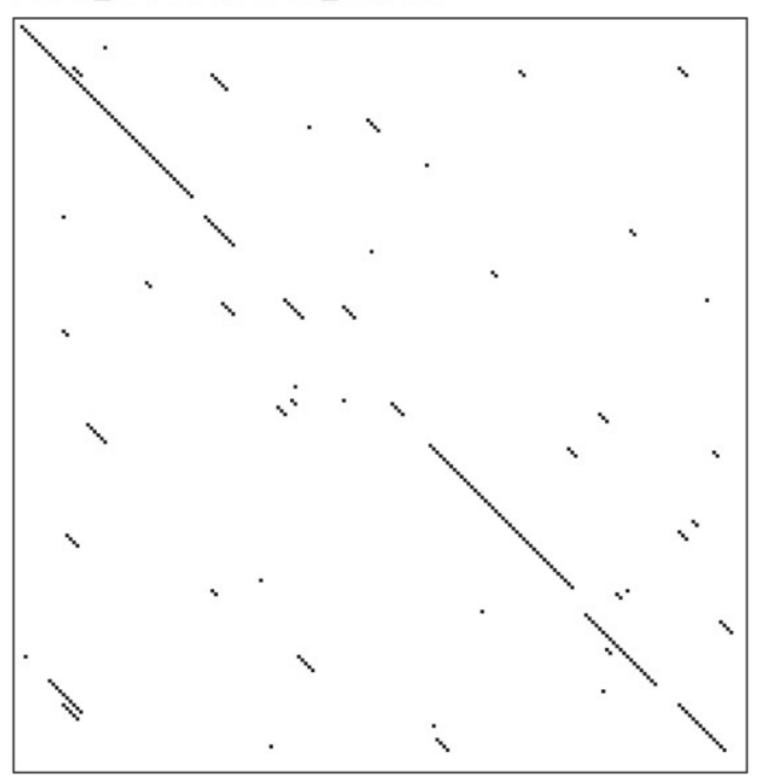


An alignment matrix

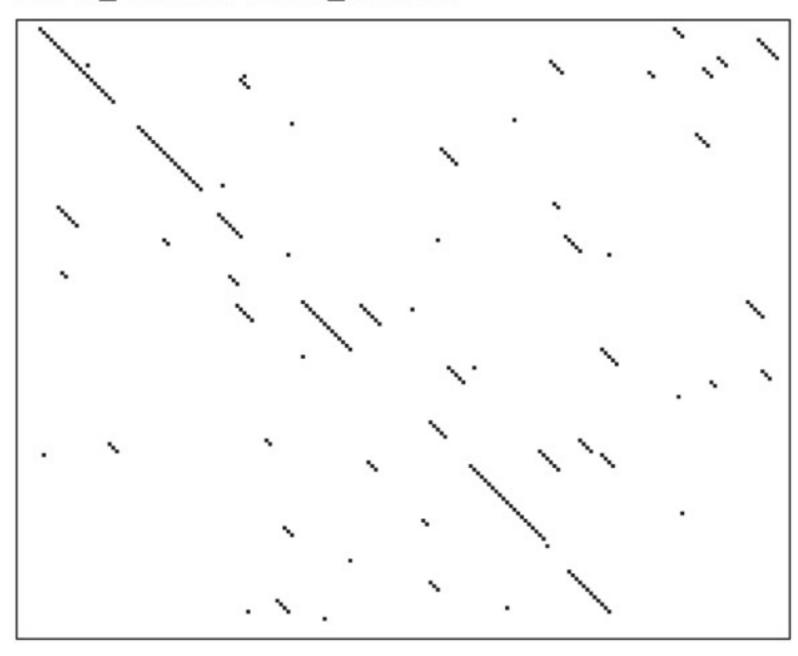




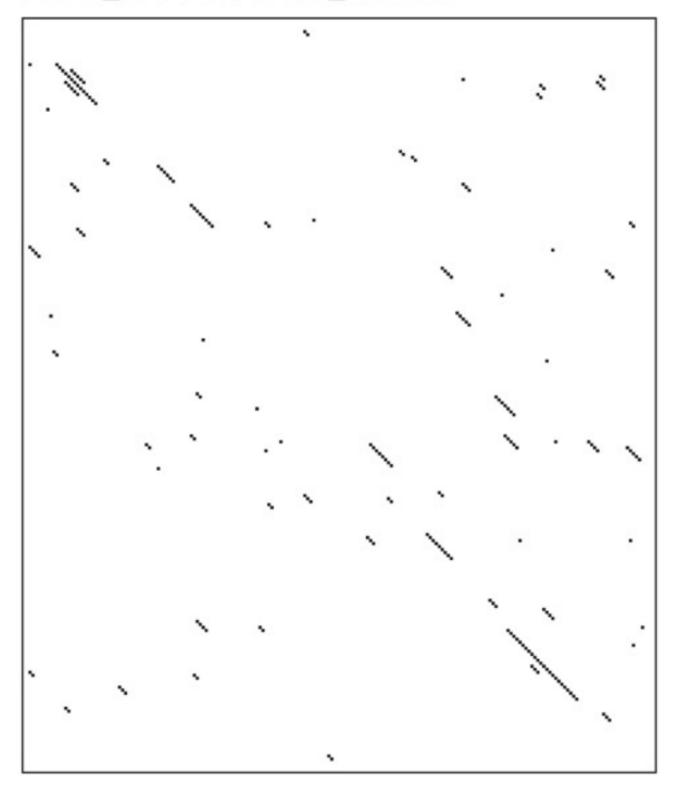
PAPA_CARPA / ACTN_ACTCH



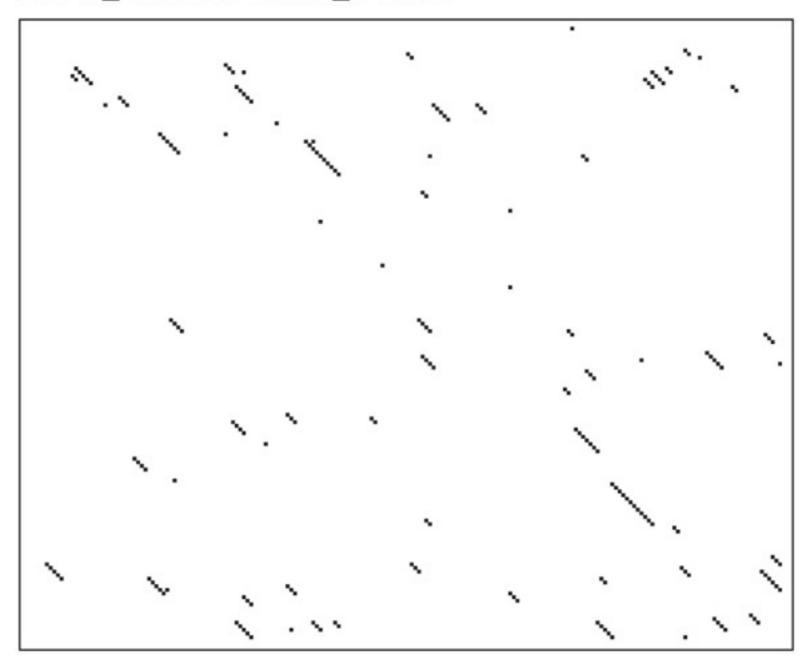
PAPA_CARPA / CATL_HUMAN

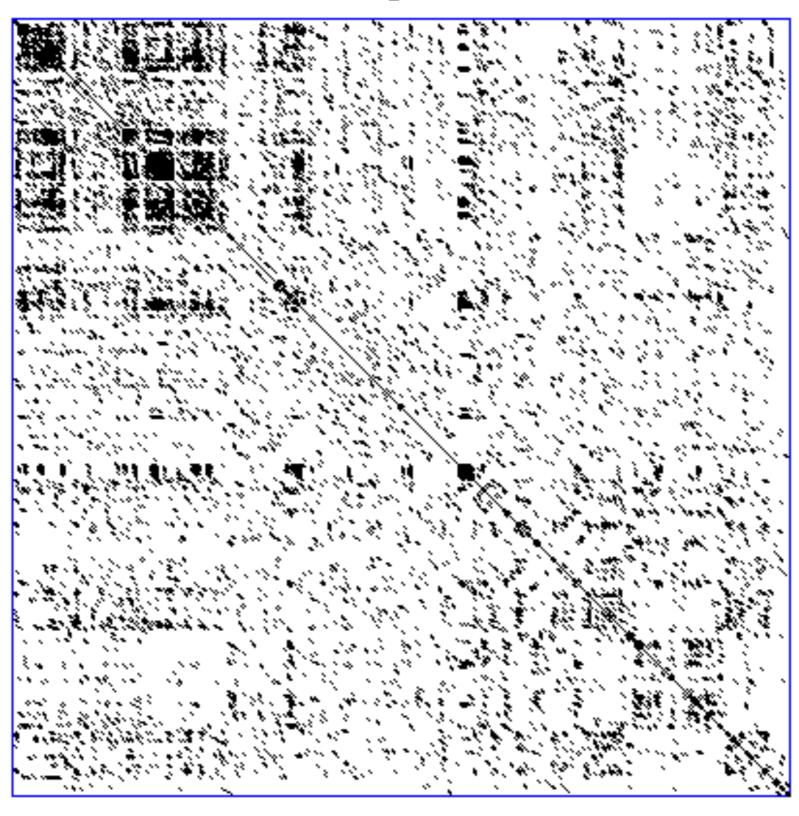


PAPA_CARPA / CATB_HUMAN



PAPA_CARPA / STPA_STAAU





(A) local PI3-kinase DRHNSNIMVKDDGQLFHIDFG CAMPPK DLKPENLLIDQQGYIQVTDFG (B) global PI3-kinase HQLGNLR--LEECRI---MSSAKRPLWLNWENPDIMSELLFQNNEIIFKNGDDLRQDMLT camppk gnaaaakkgxeqesvkeflakakedflkkwenpaqntahldqferiktlgtgsfgrvml-110 PI3-kinase LQIIRIME--NIWQNQGLDLRMLPYGCLSIGDCVGLIEVVRNSHTIMQ-IQCKGGLKGAL CAMP PK ---VKHMETGNHYAMKILDKQKVVK-----LKQIEHTLNEKRILQAVNFPFLVKLEF 130 PI3-kinase QFNSHT-LHQWLKDKNKGEIYDAA--IDLFTRSCAGYCVATFILGIGDRHNSNIMVKD-D camppk sfkdnsnlymvmeyvpggemfshlrrigrfsepharfyaaqivltfeylhsldliyrdlk

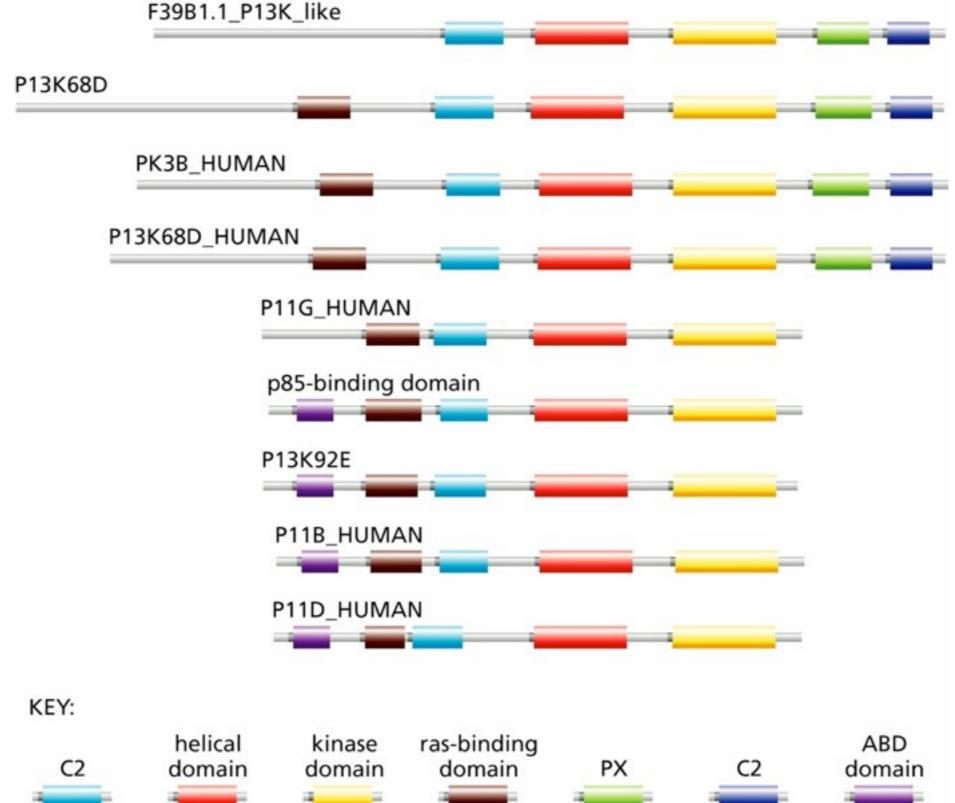
PI3-kinase GQLFHIDFGHFLDHKKKKFGYKRERVP----FVLTQDFL---IVISKGAQECTKTREFE camppk penulidaagyi--avtdfgfak-rvkgrtwxlcgtpeylapeiilskgynkavdwwalg 170

230 240 250 PI3-kinase RF-QEMC--YKAYLAIRQHANLFINLFSMMLGSGMPELQSFDDIAYIRKTLALDKTEQEA CAMP PK VLIYEMAAGYPPFFA-DQPIQIYEKIVSGKVR--FPSHFSSDLKDLLRNLLQVDLTKR--230 280

PI3-kinase LEYFMKQMNDAHHGGWTTKMDWI--290

Global FTFTALILLAVAV F--TAL-LLA-AV

Local FTFTALILL-AVAV
--FTAL-LLAAV--



Inserting gaps

(A) Bovine PI-3Kinase p110a cAMP-dependent protein kinase	LN <mark>wenp</mark> dimsel <mark>lfqnnei</mark> ifkngddlrqd <mark>ml</mark> tlqiirimeniwqnqgldlrmlpygclsigdcvgliev <mark>v</mark> rnshtimqiqckgglkgal <mark>wenp</mark> aqntah <mark>l</mark> dqfer <mark>i</mark> ktlgtgsfgrv <mark>ml</mark> vkhmetgnhyamkildkqkvvklkqiehtlnekrilqa <mark>v</mark> nfpflvklefsfkdnsnly
Bovine PI-3Kinase p110a cAMP-dependent protein kinase	QFNSHTLHQWLKDKNKGEIYDAAIDLFTRSCAGYCVATFILGIGDRHNSNIMVKDDGQLFHIDFGHFLDHK <mark>k</mark> kkf <mark>g</mark> ykrervpfvltqdf MVMEYVPGGEMFSHLRRIGRFSEPHARFYAAQIVLTFEYLHSLDLIYRDLKPENLLIDQQGYIQVTDFGFA <mark>K</mark> RVK <mark>G</mark> RTWXLCGTPEYLAP
Bovine PI-3Kinase p110a cAMP-dependent protein kinase	L <mark>I</mark> VI <mark>SKG</mark> AQECTKTREFERFQEMCYKAYLAIRQHANLF <mark>I</mark> NLFSMMLGSGM <mark>P</mark> ELQ <mark>SFD</mark> DIAYI <mark>R</mark> KTLALDKTEQEALEYFMKQMNDA <mark>H</mark> HGG E <mark>I</mark> IL <mark>SKG</mark> YNKAVDWWALGVLIYEMAAGYPPFFADQPIQ <mark>I</mark> YEKIVSGKVRF <mark>P</mark> SHF <mark>SSD</mark> LKDLLRNLLQVDLTKRFGNLKNGVNDIKN <mark>H</mark> KWF
Bovine PI-3Kinase p110a cAMP-dependent protein kinase	WTTKMDWIFHTIKQHALNATTDWIAIYQRKVEAPFIPKFKGPGDTSNFDDYEEEEIRVXINEKCGKEFSEF
(B) Bovine PI-3Kinase p110a cAMP-dependent protein kinase	LN <mark>WENP</mark> DIMSEL <mark>L FQNNEI</mark> IFKNGDDLRQDMLTLQIIRIM <mark>E</mark> NIWQNQ <mark>GL</mark> DLRMLPYGCL <mark>S</mark> IGDCVGLIEVVRNSHTIMQIQCKGGLKGAL ?- <mark>WENP</mark> AQNTAH <mark>L</mark> DQFERIKTLGTGSFGR <mark>VM</mark> LVKHMET <mark>G</mark> NHYAMK <mark>IL</mark> DKQKV-VKLKQIEHTLNEKRILQAVNFPFLVKLEFSFKDN-

Bovine PI-3Kinase p110a LNWENPDIMSELLFQNNEIIFKNGDDLRQDMLTLQIIRIMENIWQNQGLDLRMLPYGCLSIGDCVGLIEVVRNSHTIMQIQCKGGLKGAL CAMP-dependent protein kinase ?-WENPAQNTAHLDQFERIKTLGTGSFGRVMLVKHM--ETGNHYAMKILDKQKV-VKLKQIEHTLNEKRILQAVNFPFLVKLEFSFKDN
Bovine PI-3Kinase p110a QFNSHTLHQWLKDKNKGEIYDAAIDLFTRSCAGYCVATFILGIGDRHNSNIMVKD-DGQLFHIDFGHFLDHKKKKFGYKRERVPFVL--T CAMP-dependent protein kinase -SNLYMVMEYVPGGEMFSHLRR-IGRFSEPHARFYAAQIVLTFEYLHSLDLIYRDLKPENLLIDQQGYIQVTDFGFAKRVKGRTWXLCGT

Bovine PI-3Kinase p110a QDFL---IVISKGAQECTKTREFERF-QEMC--YKAYLAIRQHANLFINLFSMMLGSGMPELQSFDDIAYIRKTLALDKTEQEALEYFMK CAMP-dependent protein kinase PEYLAPEIILSKGYNKAVDWWALGVLIYEMAAGYPPFFA-DQPIQIYEKIVSGKVRF--PSHFSSDLKDLLRNLLQVDLTKR--FGNLKN

What is an optimal alignment?

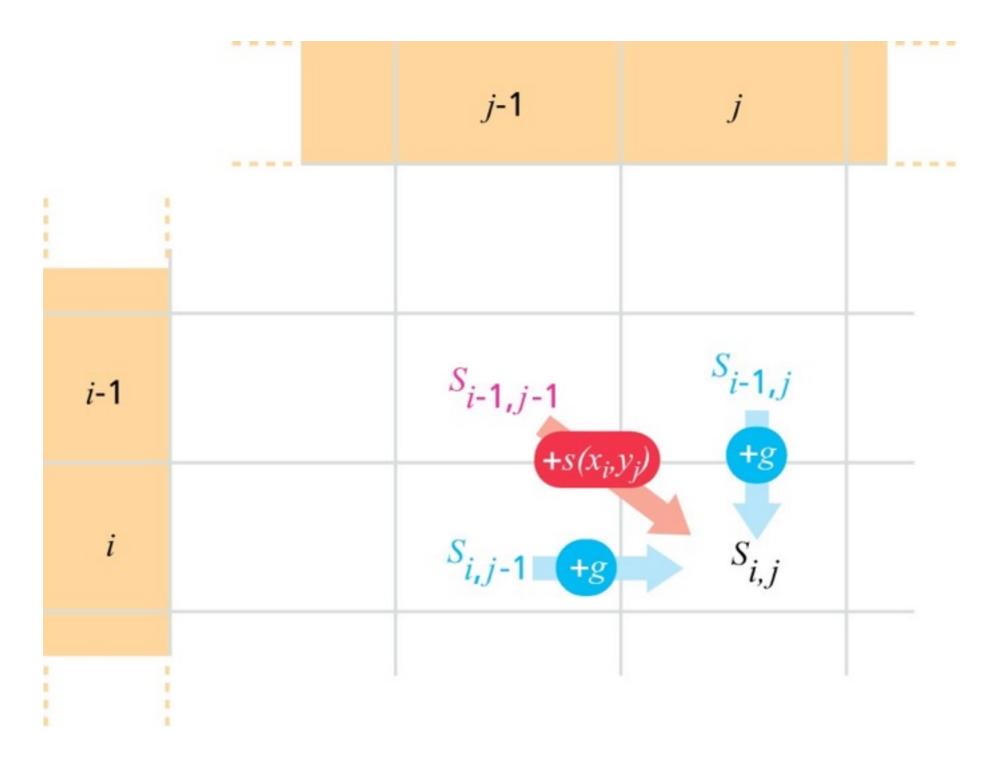
```
THISSEQUENCE
T H A T S E Q U E N C E
T H A T S E Q U E N C E
      | 4/12 Identical
THISISASEQUENCE
THISISA-SEQUENCE
       TH---ATSEQUENCE
```

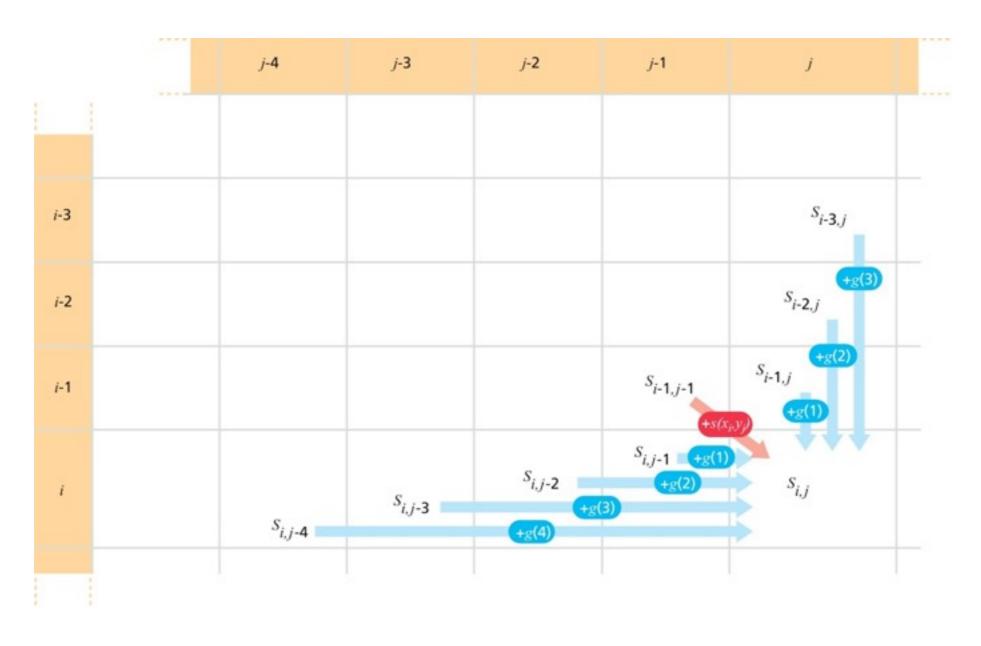
Different scoring

```
THISSEQUENCE
5 8-1 1 4 5 6 0 5 6 9 5 Score = 52
T H A T S E Q U E N C E
THATSEQUENCE
5 \ 8-1-1-2 \ 0-1 \ 0 \ 5 \ 0 \ 0 \ 5 Score = 18
THISISASEQUENCE
 HISISA - SEQUENCE
5 8 0 0 0 0 4 0 4 5 6 0 5 6 9 5 Score = 56
TH----ATSEQUENCE
```

With Gap cost

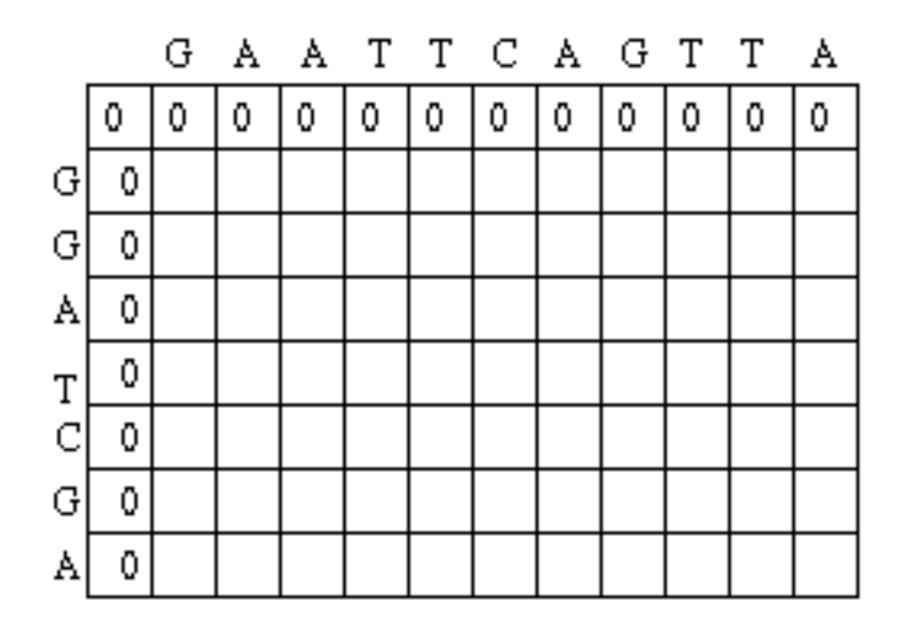
```
THISSEQUENCE
5 8-1 1 4 5 6 0 5 6 9 5 Score = 52
T H A T S E Q U E N C E
THATSEQUENCE
5 \ 8-1-1-2 \ 0-1 \ 0 \ 5 \ 0 \ 0 \ 5 Score = 18
THISISASEQUENCE
THISISA-SEQUENCE
5 8-1-1-1-1 4-1 4 5 6 0 5 6 9 5 Score = 51
TH---ATSEQUENCE
```





	<i>j</i> -4	<i>j</i> -3	<i>j</i> -2	<i>j</i> -1	j	
<i>i</i> -3				S _{i-3,j-1}		
i-2				S _{i-2,j-1}		
<i>i</i> -1	S _{i-1,j-4}	S _{i-1,j-3}	S _{i-1,j-2}	S _{i-1,j-1}		
i					$S_{i,j}$	

Initialisation step: Create Matrix with M + 1 columns and N + 1 rows. First row and column filled with 0.

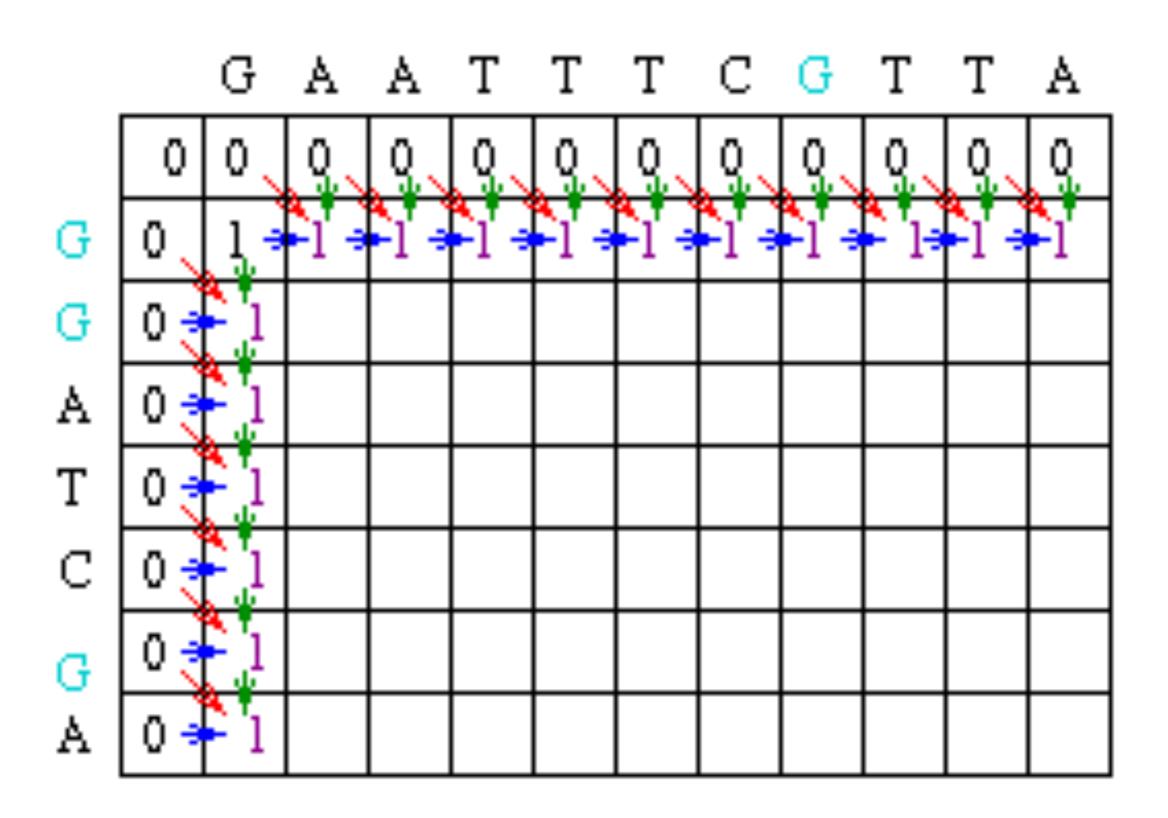


Matrix fill step: Each position $M_{i,j}$ is defined to be the MAXIMUM score at position i,j $M_{i,i} = MAXIMUM$ [

 $M_{i-1, j-1} + s_{i,,j}$ (match or mismatch in the diagonal) $M_{i, j-1} + w$ (gap in sequence #1) $M_{i-1, j} + w$ (gap in sequence #2)]

		G	A	A	Т	Т	С	A	G	Т	Т	A
	ó	ę.	0	0	0	0	0	0	0	0	0	0
G	0	1										
G	0											
A	0											
Т	0											
С	0											
G	0											
A	0											

Fill in rest of row 1 and column 1



Fill in column 2

		G	A	A	T	Т	С	Α	G	Т	T	A
	0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	Ţ	1	1	1	1	1	1	1	1	1
G	0	1 =	j									
A	0	1 =	2									
Т	0	1 =	2									
С	0	1 =	2									
G	0	1 =	2									
A	0	1 =	2									

Fill in column 3

		G	Α	A	T	Т	С	A	G	Т	Т	Α
	0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	Į	1	1	1	1	1	1	1	1
G	0	1	1									
A	0	1	2									
T	0	1	2									
С	0	1	2									
G	0	1	2 🕏									
A	0	1	2 =									

Column 3 with answers

		G	Α	A	T	Т	С	Α	G	T	T	Α
	0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	Ţ	1	1	1	1	1	1	1	1
G	0	1	1 =	ļ								
A	0	1	2	2								
T	0	1	2	2								
С	0	1	2	2								
G	0	1	2	2								
A	0	1	2 =	3								

Fill in rest of matrix with answers

		G	Α	Α	Т	Т	С.	A	G	Т	T	Α
	0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1	1
G	0	1	1	1	1	1	1	1	2	2	2	2
A	0	1	2	2	2	2	2	2	2	2	2	3
Т	0	1	2	2	3	3	3	3	3	3	3	3
С	0	1	2	2	3	3	3	4	4	4	4	4
G	0	1	2	2	3	3	3	4	4	5	5	5
A	0	1	2	3	3	3	3	4	5	5	5	6

Traceback step:
Position at current cell and look at direct predecessors

		G	Α	Α	T	T	С	Α	G	Т	T	Α
	0	0	0	0	0	0	0	0	0	0	0	0
G	0	1	1	1	1	1	1	1	1	1	1	1
G	0	1	1	1	1	1	1	1	2	2	2	2
A	0	1	1	2	2	2	2	2	2	2	2	3
Т	0	1	2	2	3	3	3	3	3	3	3	3
С	0	1	2	2	3	3	4	4	4	4	4	4
G	0	1	2	2	3	3	4	4	5	5	5	5
A	0	1	2	3	3	3	4	5	5	5	5	= 6

Traceback step:
Position at current cell and look at direct predecessors

		G	A	A	Т	Т	С	A	G	Т	Т	A
	0	0	0	0	0	0	0	0	0	0	0	
G	0	1	1	1	1	1	1	1	1	1	1	
G	0	1	1	1	1	1	1	1	2	2	2	
A	0	1	1	2	2	2	2	2	2	2	2	
T	0	1	2	2	3	3	3	3	3	3	3	
С	0	1	2	2	3	3	4	4	4	4	4	
G	0	1	2	2	3	3	4	4	5	5	5	
A												= 6

Traceback step:
Position at current cell and look at direct predecessors

		G	A	A	Т	Т	С	A	G	Т	T	Α
	0	0	0	0	0	0	0	0	0	0	0	
G	0	1	1	1	1	1	1	1	1	1	1	
G	0	1	1	1	1	1	1	1	2	2	2	
A	0	1	1	2	2	2	2	2	2	2	2	
Т	0	1	2	2	3	3	3	3	3	3	3	
С	0	1	2	2	3	3	4	4	4	4	4	
G	0	1	2	2	3	3	4	4	5	5	5	
A												- 6

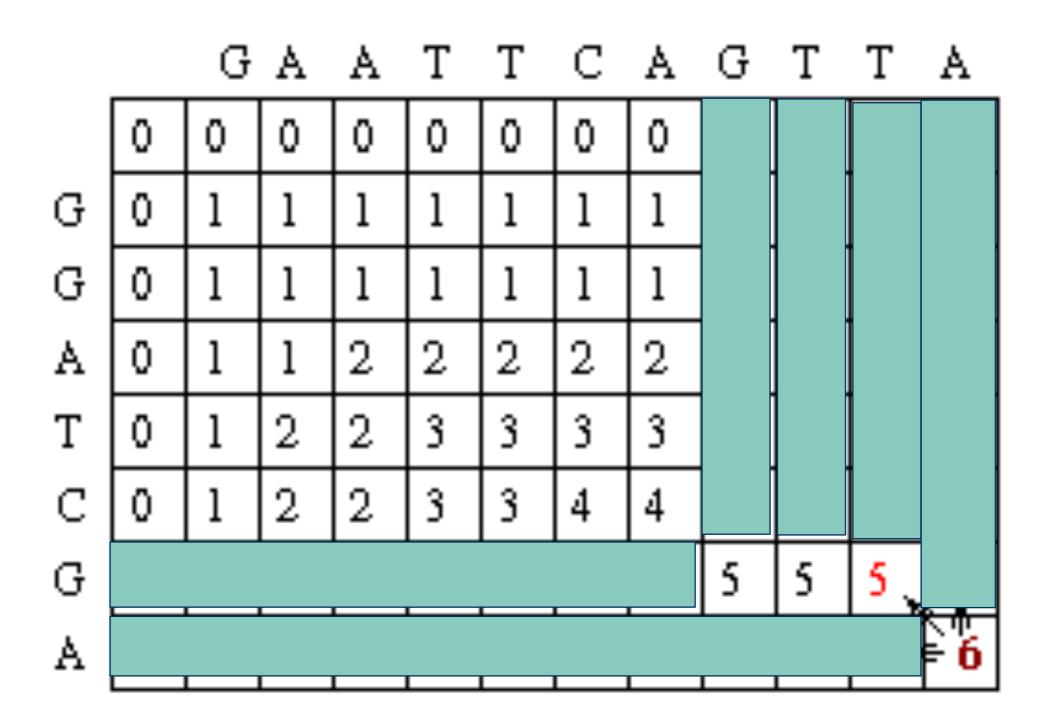
Traceback step:
Position at current cell and look at direct predecessors

		G	A	A	Т	Т	С	A	G	T	Т	Α
	0	0	0	0	0	0	0	0	0	0		
G	0	1	1	1	1	1	1	1	1	1		
G	0	1	1	1	1	1	1	1	2	2		
A	0	1	1	2	2	2	2	2	2	2		
T	0	1	2	2	3	3	3	3	3	3		
С	0	1	2	2	3	3	4	4	4	4		
G	0	1	2	2	3	3	4	4	5	5	5	
A												= 6

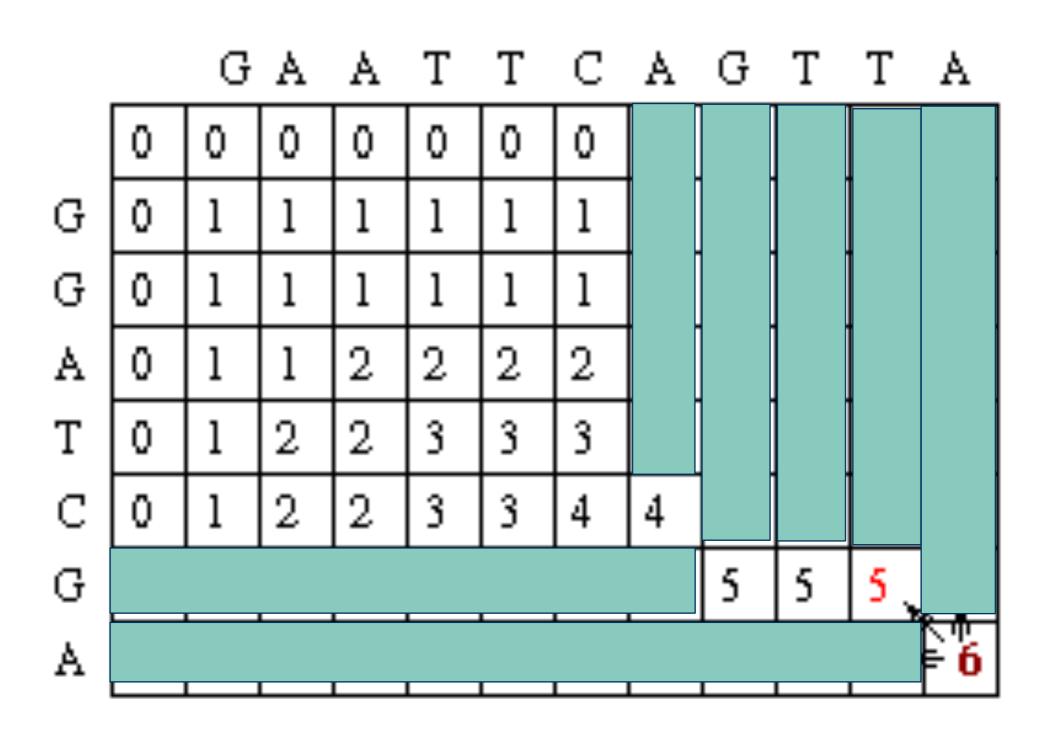
Traceback step:
Position at current cell and look at direct predecessors

		G	A	A	T	T	С	A	G	T	T	Α
	0	0	0	0	0	0	0	0	0			
G	0	1	1	1	1	1	1	1	1			
G	0	1	1	1	1	1	1	1	2			
A	0	1	1	2	2	2	2	2	2			
Т	0	1	2	2	3	3	3	3	3			
С	0	1	2	2	3	3	4	4	4			
G	0	1	2	2	3	3	4	4	5	5	5	
A												= 6

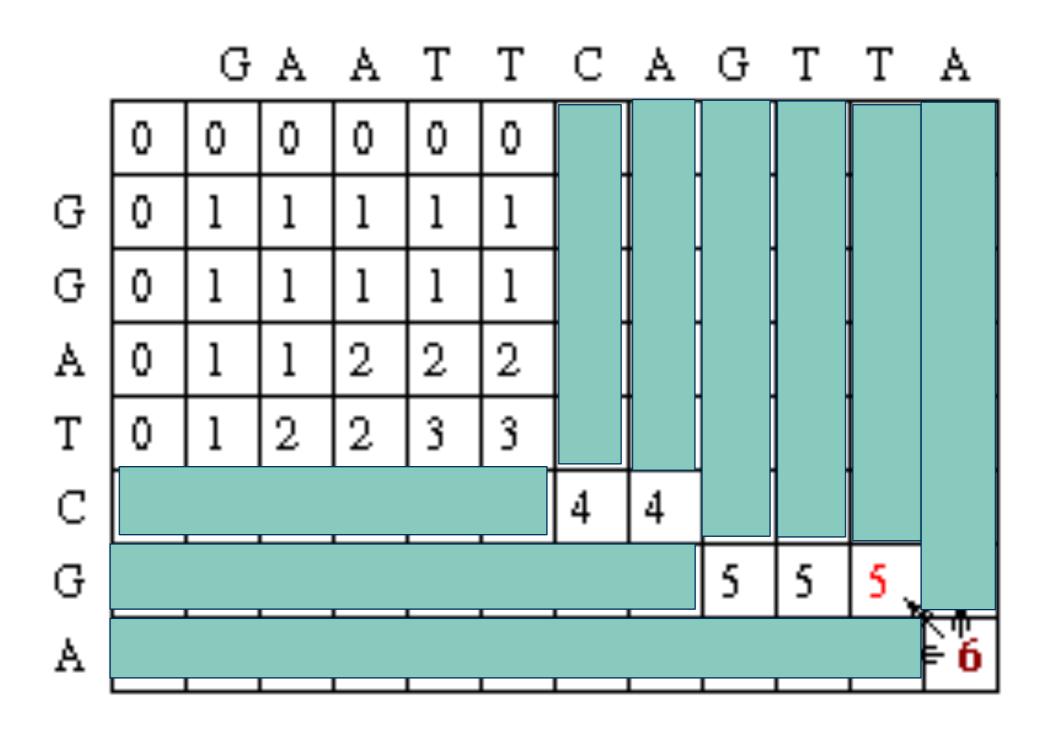
Traceback step:
Position at current cell and look at direct predecessors



Traceback step:
Position at current cell and look at direct predecessors



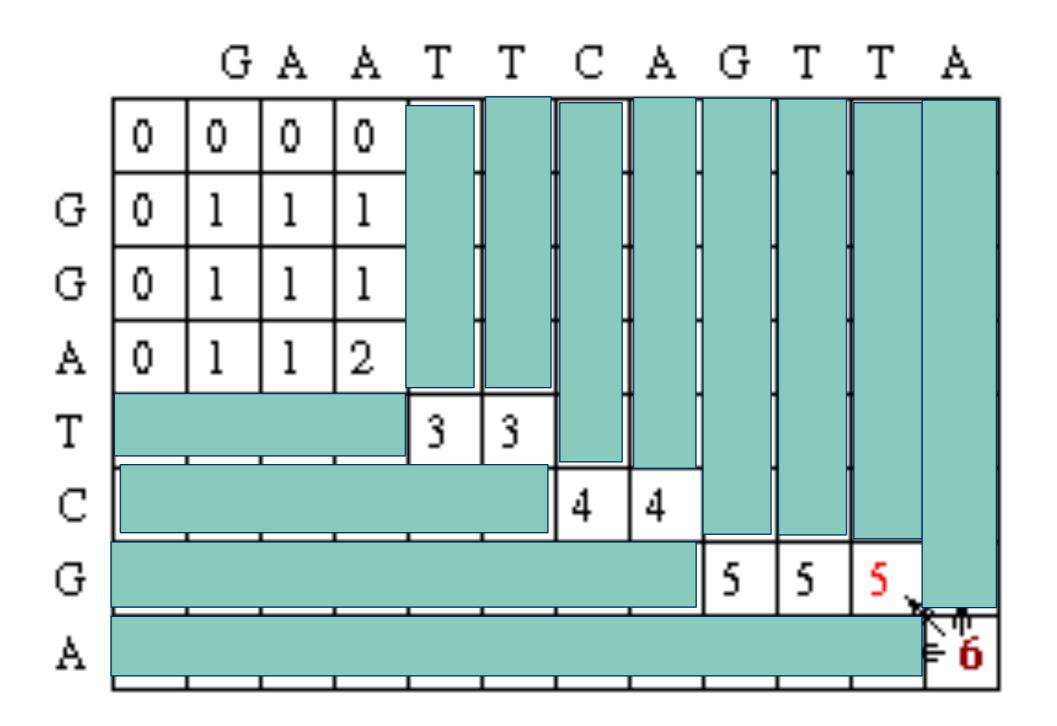
Traceback step:
Position at current cell and look at direct predecessors



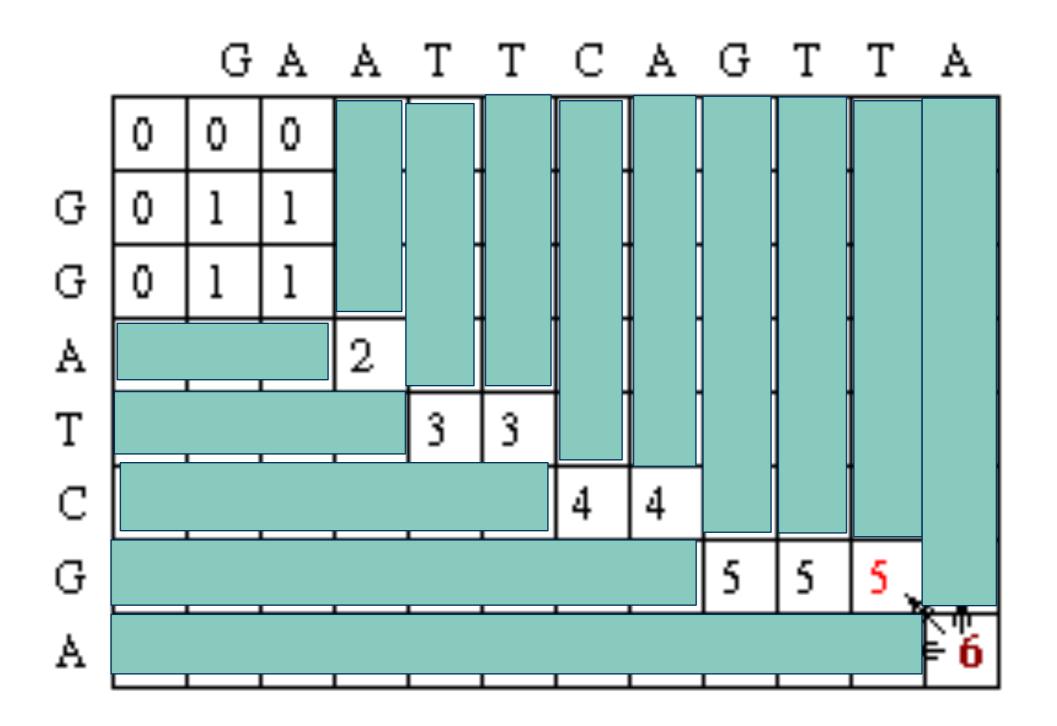
Traceback step:
Position at current cell and look at direct predecessors

GAATTCAGTTA G G G

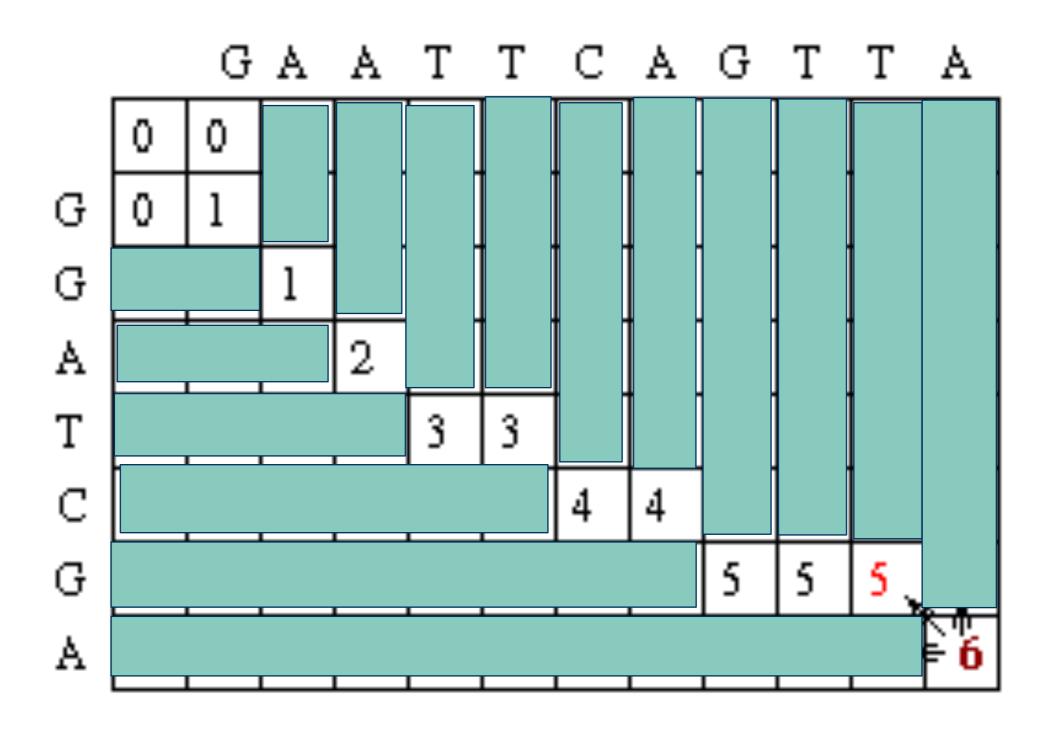
Traceback step:



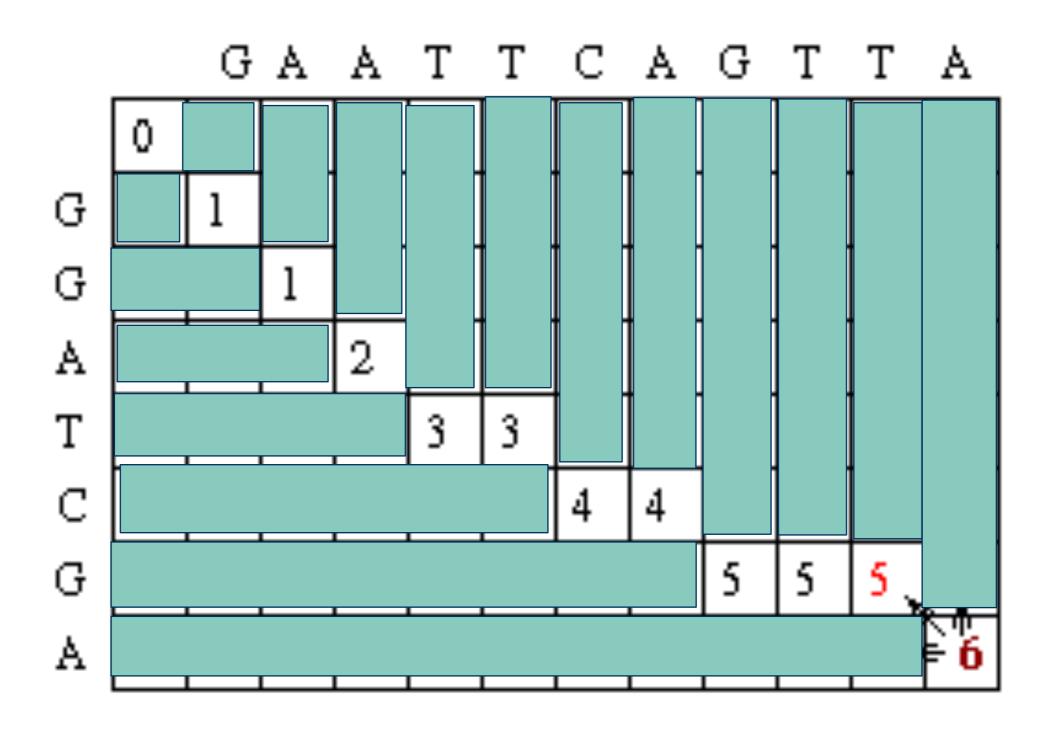
Traceback step:



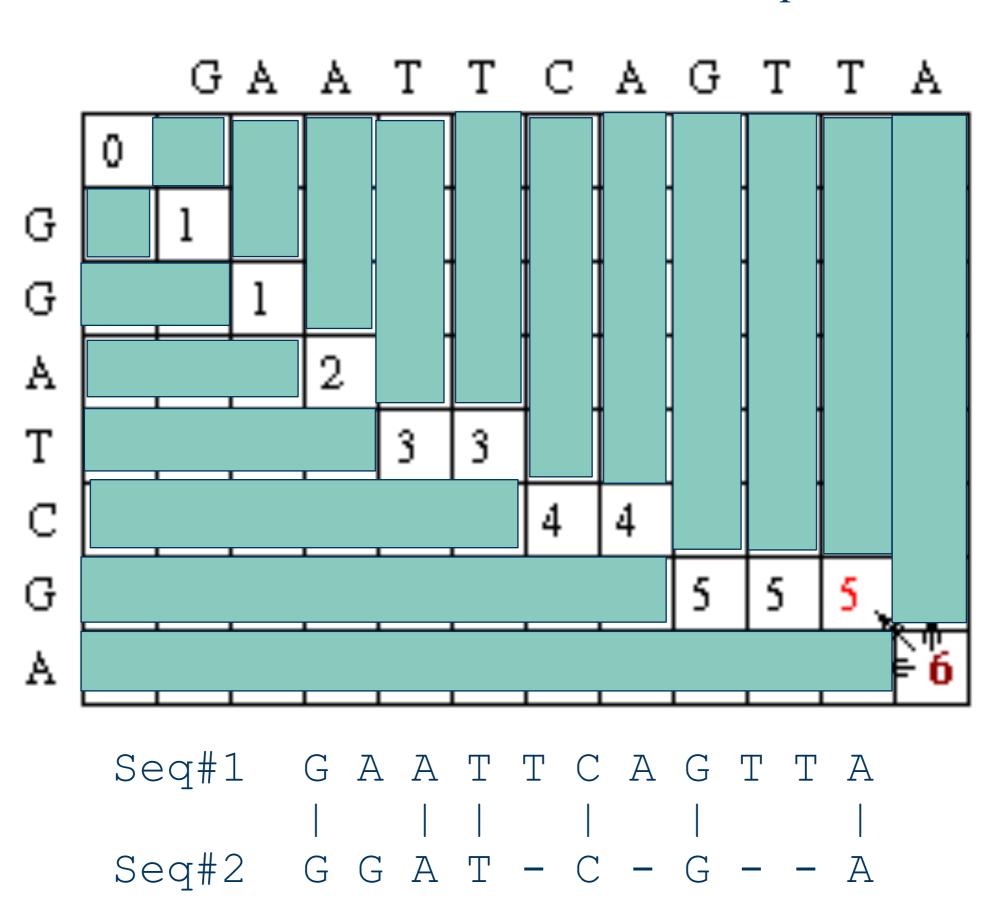
Traceback step:



Traceback step:



Traceback step:



Pseudocode

```
for i=0 to length(A)
  F(i,0) \leftarrow d*i
for j=0 to length(B)
  F(0,j) \leftarrow d*j
for i=1 to length(A)
  for j=1 to length(B)
     Match \leftarrow F(i-1,j-1) + S(A<sub>i</sub>, B<sub>j</sub>)
     Delete \leftarrow F(i-1, j) + d
     Insert \leftarrow F(i, j-1) + d
     F(i,j) \leftarrow max(Match, Insert,
Delete)
   }
```

Traceback

```
AlignmentA ← ""
                          AlignmentB ← ""
                           i \leftarrow length(A)
                            j \leftarrow length(B)
                     while (i > 0 \text{ or } j > 0)
if (i > 0 \text{ and } j > 0 \text{ and } F(i,j) == F(i-1,j-1) + S(A_i, B_j))
                    AlignmentA \leftarrow A<sub>i</sub> + AlignmentA
                    AlignmentB \leftarrow B<sub>j</sub> + AlignmentB
                                 i \leftarrow i - 1
                                 j ← j - 1
         else if (i > 0 \text{ and } F(i,j) == F(i-1,j) + d)
                    AlignmentA \leftarrow A<sub>i</sub> + AlignmentA
                   AlignmentB ← "-" + AlignmentB
                                 i \leftarrow i - 1
           else (j > 0 \text{ and } F(i,j) == F(i,j-1) + d)
                   AlignmentA ← "-" + AlignmentA
                    AlignmentB \leftarrow B<sub>j</sub> + AlignmentB
                                 j ← j - 1
```

Scoring Matrices

 $S = [s_{ij}]$ gives score of aligning character i with character j for every pair i, j.

```
C 12

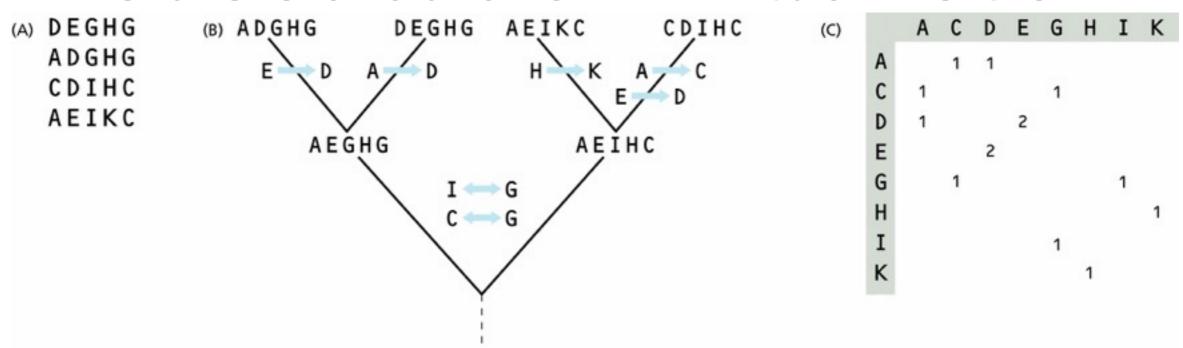
S 0 2

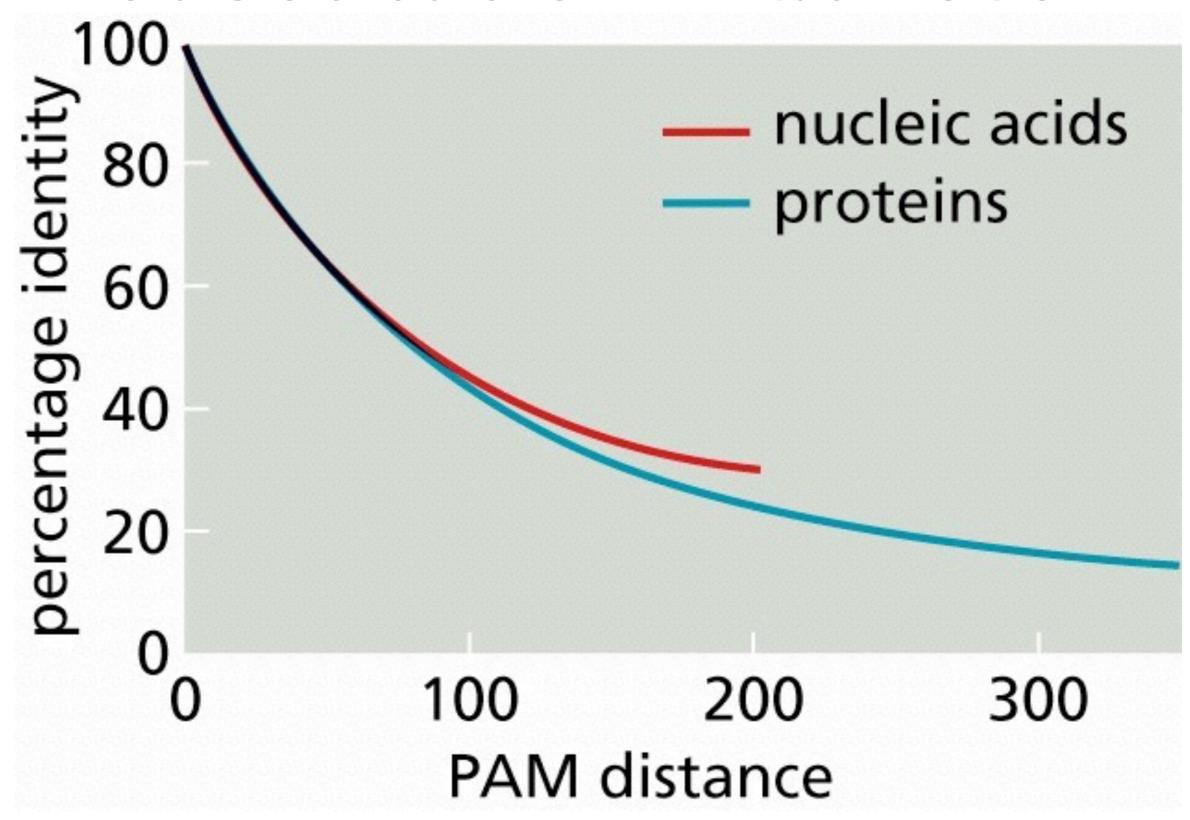
T -2 1 3

P -3 1 0 6

A -2 1 1 1 2

C S T P A = 1
```





Log Odds Ratios

$$S_{i,j} = \log \frac{p_i \cdot M_{i,j}}{p_i \cdot p_j} = \log \frac{M_{i,j}}{p_j} = \log \frac{observed\ frequency}{expected\ frequency}$$

$$PAM_n(i,j) = log \frac{f(i)M^n(i,j)}{f(i)f(j)} = log \frac{M^n(i,j)}{f(j)}$$

$$PAM_n(i,j) = log \frac{f(i)M^n(i,j)}{f(i)f(j)} = log \frac{M^n(i,j)}{f(j)}$$

The PAM Family

Define a *family* of substitution matrices — PAM 1, PAM 2, etc. — where PAM n is used to compare sequences at distance n PAM.

$$PAM n = (PAM 1)^n$$

Do not confuse with scoring matrices!

Scoring matrices are derived from PAM matrices to yield log-odds scores.

$$PAM_n(i,j) = log \frac{f(i)M^n(i,j)}{f(i)f(j)} = log \frac{M^n(i,j)}{f(j)}$$

PAM matrices

Let M be a PAM 1 matrix. Then,

$$\sum_{i} p_{i}(1-M_{ii})=0.01$$

• Reason: M_{ii} s are the probabilities that a given amino acid does not change, so (1- M_{ii}) is the probability of mutating away from i.

$$PAM_n(i,j) = log \frac{f(i)M^n(i,j)}{f(i)f(j)} = log \frac{M^n(i,j)}{f(j)}$$

```
P -3 -1 -1 7
A 0 1 0 -1 4
G -3 0 -2 -2 0 6
N -3 1 0 -2 -2 0 6
D -3 0 -1 -1 -2 -1 1 6
E -4 0 -1 -1 -1 -2 0 2 5
Q -3 0 -1 -1 -1 -2 0 0 2 5
H -3 -1 -2 -2 -2 -2 1 -1 0 0 8
R -3 -1 -1 -2 -1 -2 0 -2 0 1 0 5
K -3 0 -1 -1 -1 -2 0 -1 1 1 -1 2 5
M -1 -1 -1 -2 -1 -3 -2 -3 -2 0 -2 -1 -1 5
I -1 -2 -1 -3 -1 -4 -3 -3 -3 -3 -3 -3 -3 1 4
L -1 -2 -1 -3 -1 -4 -3 -4 -3 -2 -3 -2 -2 2 2 4
V -1 -2 0 -2 0 -3 -3 -3 -2 -2 -3 -3 -2 1 3 1 4
F -2 -2 -2 -4 -2 -3 -3 -3 -3 -1 -3 -3 0 0 0 -1 6
Y -2 -2 -2 -3 -2 -3 -2 -3 -2 -1 2 -2 -2 -1 -1 -1 -1 3 7
W -2 -3 -2 -4 -3 -2 -4 -4 -3 -2 -2 -3 -3 -1 -3 -2 -3 1 2 11
  CSTPAGNDEOHRKMILVFYW
```

(A)		12345
	1	ATCKQ
	2	ATCRN
	3	ASCKN
	4	SSCRN
	5	SDCEQ
	6	SECEN
	7	TECRQ

(B)	q_{QN}	$q_{ m NN}$	q_{QQ}	$p_{\rm N}$	p_{Q}
C=62%	0.114	0.057	0.029	0.114	0.086
C=50%	0.117	0.025	0.058	0.084	0.117
C = 40%	_	_	-	-	-

Equivalent PAM and Blossum matrices (according to H)

- PAM100 ==> Blosum90
- PAM120 ==> Blosum80
- PAM160 ==> Blosum60
- PAM200 ==> Blosum52
- PAM250 ==> Blosum45

```
C 11
```

Difference between Pam and Blosum

- PAM matrices are based on an explicit evolutionary model (i.e. replacements are counted on the branches of a phylogenetic tree), whereas the BLOSUM matrices are based on an implicit model of evolution.
- The PAM matrices are based on mutations observed throughout a global alignment, this includes both highly conserved and highly mutable regions. The BLOSUM matrices are based only on highly conserved regions in series of alignments forbidden to contain gaps.
- The method used to count the replacements is different: unlike the PAM matrix, the BLOSUM procedure uses groups of sequences within which not all mutations are counted the same.
- Higher numbers in the PAM matrix naming scheme denote larger evolutionary distance, while larger numbers in the BLOSUM matrix naming scheme denote higher sequence similarity and therefore smaller evolutionary distance. Example: PAMI50 is used for more distant sequences than PAMI00; BLOSUM62 is used for closer sequences than BLOSUM50.

Nucleotide Matrices

Dayhoff's PAM matrix

	A	R	N	D	C
A	9867	2	9	10	3
R	1	9913	1	0	1
N	4	1	9822	36	0
D	6	0	42	9859	0
C	1	1	0	0	9973

All entries × 104

-20

-96

67

(A)

-117

(B)

Gap models

- Gap-extension
- Gap opening cost

Local and global

(A) local PI3-kinase DRHNSNIMVKDDGQLFHIDFG cAMP PK DLKPENLLIDQQGYIQVTDFG (B) global PI3-kinase HQLGNLR--LEECRI---MSSAKRPLWLNWENPDIMSELLFQNNEIIFKNGDDLRQDMLT CAMP PK GNAAAAKKGXEQESVKEFLAKAKEDFLKKWENPAQNTAHLDQFERIKTLGTGSFGRVML-PI3-kinase LQIIRIME--NIWQNQGLDLRMLPYGCLSIGDCVGLIEVVRNSHTIMQ-IQCKGGLKGAL ---VKHMETGNHYAMKILDKQKVVK-----LKQIEHTLNEKRILQAVNFPFLVKLEF PI3-kinase QFNSHT-LHQWLKDKNKGEIYDAA--IDLFTRSCAGYCVATFILGIGDRHNSNIMVKD-D CAMP PK SFKDNSNLYMVMEYVPGGEMFSHLRRIGRFSEPHARFYAAQIVLTFEYLHSLDLIYRDLK PI3-kinase GQLFHIDFGHFLDHKKKKFGYKRERVP----FVLTQDFL---IVISKGAQECTKTREFE CAMPPK PENLLIDQQGYI--QVTDFGFAK-RVKGRTWXLCGTPEYLAPEIILSKGYNKAVDWWALG

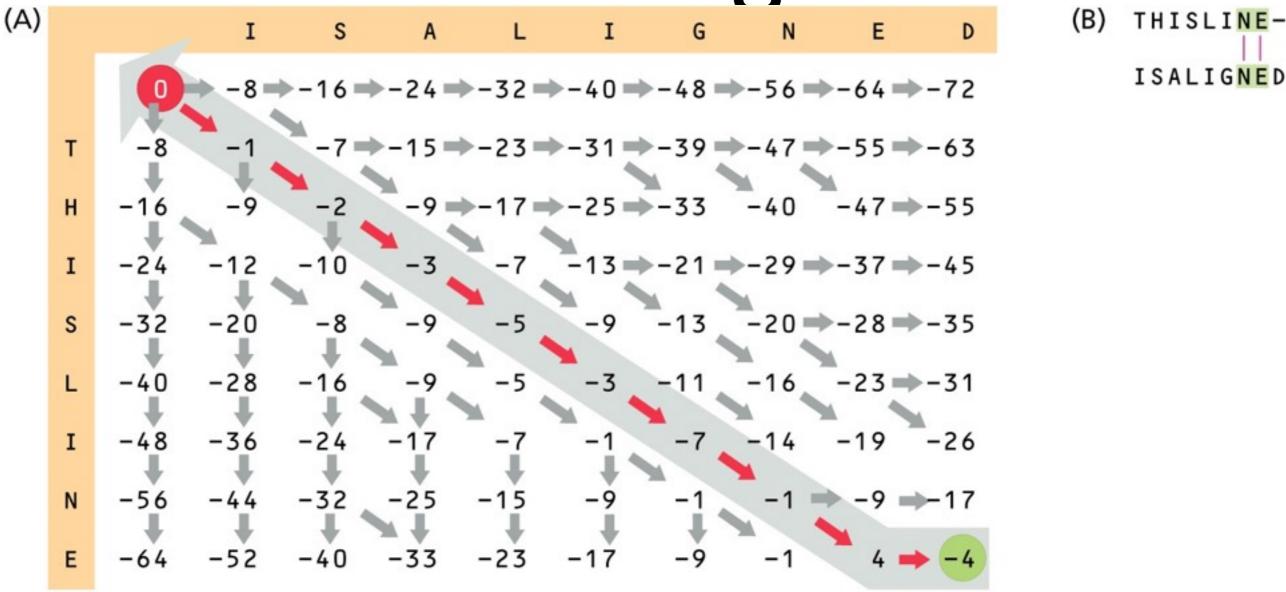
CAMPPK VLIYEMAAGYPPFFA-DQPIQIYEKIVSGKVR--FPSHFSSDLKDLLRNLLQVDLTKR--

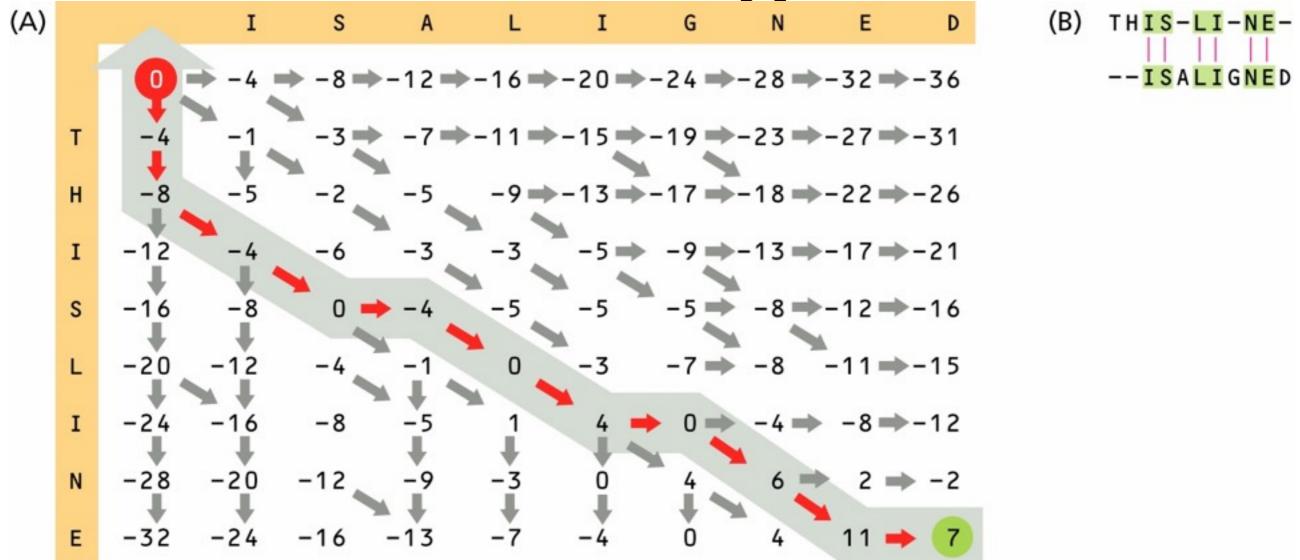
Global alignment Needleman-Wunch

Local alignment Smith Waterman

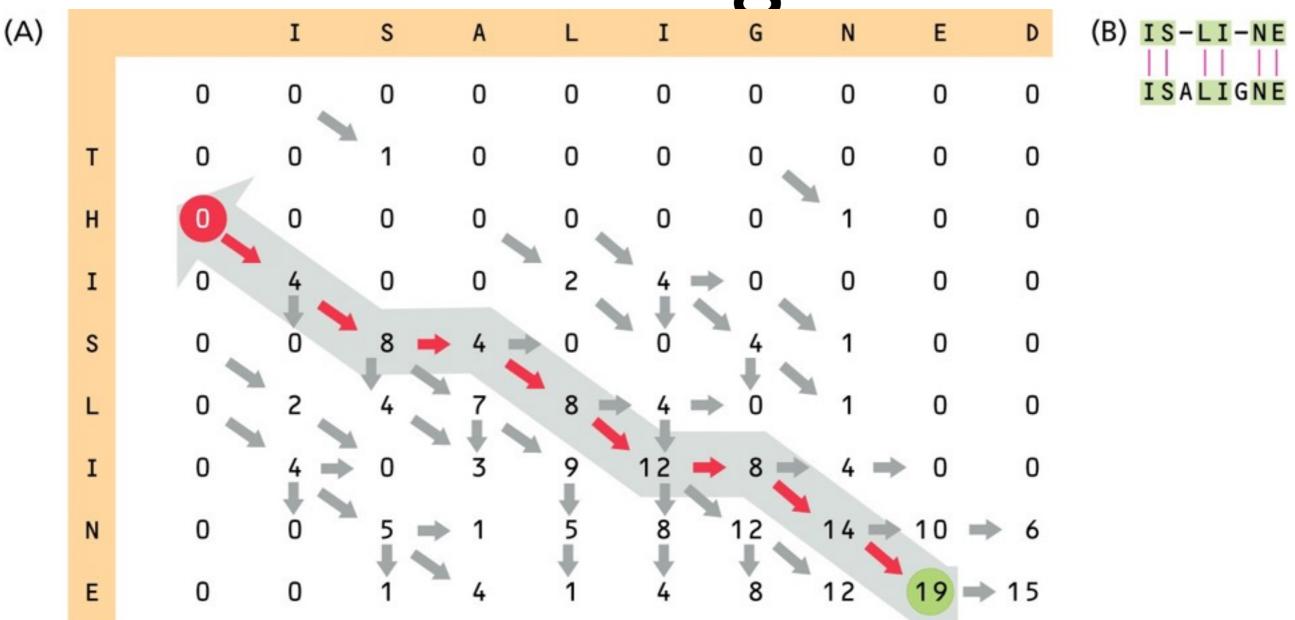
	GAP	М	N	Α	L	s	D	R	Т
GAP	0	-12	-16	-20	-24	-28	-32	-36	-40
М	-12	6 (6)	-6	-10	-14	-18	-22	-26	-30
G	-16	-6 -6	6(0)	-5	-10	-13	-17	-22	-26
s	-20	-10	-5	7	-5	-8 、	-13	-17	-21
D	-24	-14	-8	-5	3	-5	`-4 、	-14	-17
R	-28	-18	-14	-9	`-8 <u>`</u>	3	-6	`2 、	-10
Т	-32	-22	-18	-13	-11	-7	3	-7	` 5 ¦
Т	-36	-26	-22	-17	-15	-10	-7	2	-4
Е	-40	-30	-25	-21	-20	-15	-7	-8	2
Т	-44	-34	-30	-24	-23	-19	-15	-8	\ ₋₅ ¦

	GAP	М	N	Α	L	s	D	R	Т
GAP	0	0	0	0	0	0	0	0	0
М	0	6	0	0	4	0	0	0	0
G	0	0	6	1	0	5	1	0	0
S	0	0	1	7	0	2	5	1	1
D	0	0	2	1	3	0	6	4	1
R	0	0	0	0	0	3	0	12	3
Т	0	0	0	1	0	1	3	0	15
Т	0	0	0	1	0	1	1	2	3
Е	0	0	1	0	0	0	4	0	2
Т	0	0	0	2	0	1	0	3	3

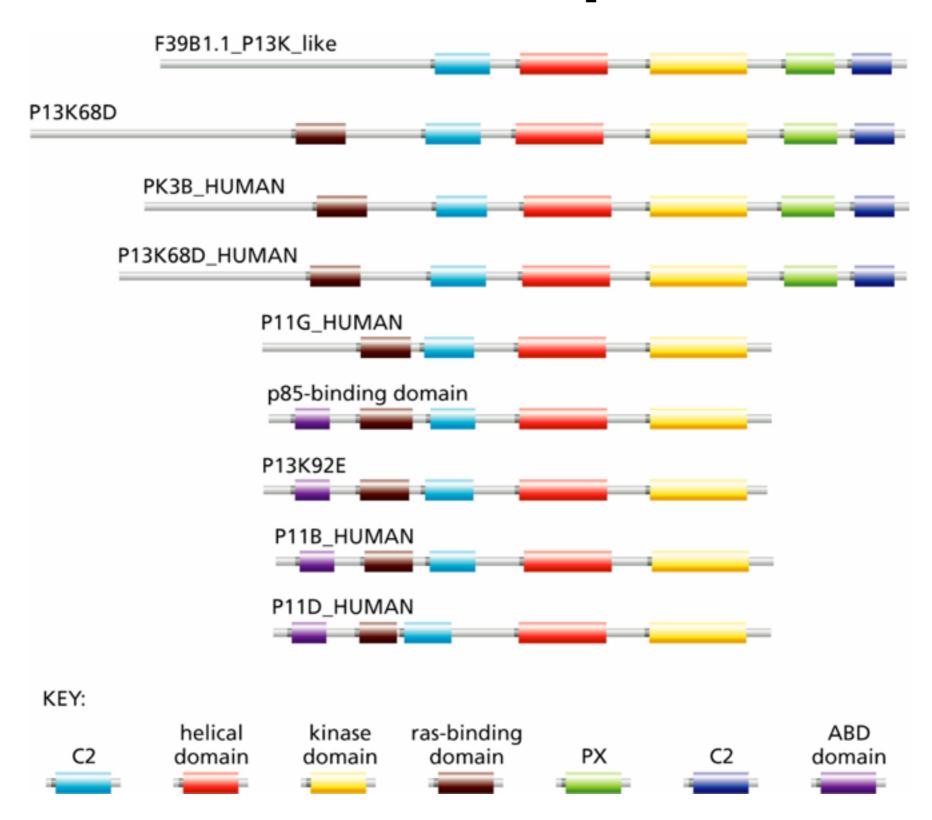




(A)			I	S	Α	L	I	G	N	E	D (B)	SLI-NE
		0	0	0	0	0	0	0	0	0	0	ALIGNE
	Т	0	0	1	0	0	0	0	0	0	0	
	н	0	0	0	0	0	0	0	1	0	0	
	I	0	4	0	0	2	4	0	0	0	0	
	S	0	0	8	1	0	0	4	1	0	0	
	L	0	2	0	7	5	2	0	1	0	0	
	I	0	4	0	0	9	9 🗪	1	0	0	0	
	N	0	0	5	0	1	6	9	7	0	1	
	Ε	0	0	0	4	0	0	4	9	12 →	4	



Multidomain proteins



Next lecture

- O(nm) is too slow. How to speed up
- When is a "score" significant.