

BCB 5200 Introduction to Bioinformatics

Multiple Sequence Alignment

Bioinformatics and Computational Biology
Saint Louis University

Outline

- Multiple sequence alignment (MSA)
- Scoring MSA
- MSA algorithms
 - Exact approach - Dynamic programming
 - Progressive alignments - ClustalW
 - Iterative approach - MUSCLE
 - Profile-based approach - Promals
- Warning

- One of the most important contributions of biological sequences to biology and evolution is the discovery that sequences of different organisms are often related.
- They are called homologies

Multiple sequence alignment: definition

- A collection of three or more protein (or nucleic acid) sequences that are partially or completely aligned
- A model
- Indicates relationship between residues of different sequences (similarity/dissimilarity)

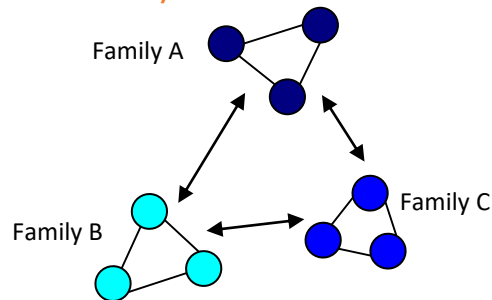
Multiple sequence alignment

```
BSUB00  RMAHYDSLTDLPNRRHAISHLTKVLNREHSLHYNTVVFFLDLNRFKVINDAL
ECU738  VMSTRDGMTGVYNRRHWETMLRNEFDNCRRHNRDATLLIIDIDHFKSINDTW
D90790  HEVGMDVLTKLLNRRFLPTIFKREIAHANRTGTPLSVLIIDVDKFKEINDTW
SYCSLL  QISSLDALTQVGNRYLFDSTLEREWQRLQRIREPLALLLCDVDFFKGFNDNY
ECAE00  NIAHRDPLTNIFNRNYFFNEL--TVQSASAQKTPYCVIMDIDHFKKVNDTW
AF0348  QAANVDSLTLGLANRAAYNAHM-ERLTAADAPS--IGLLIIDVDRLKQVNDIL
D90796  IRSNMDVLTGLPGRRVLDESFDHQLRNAEPLN--LYLMLLDIDRFKLVNDTY
Y4LL_R  HMARHDAITGLPNRQFLREEF-ERLSDHIAPSTRLAILCLDLGFKAINDAY
Y07I_M  YLADHDDLTLGLHNRRALLQHLDQRLAPGQPGP--VAALFLDLRLKAINDYL
.....
```

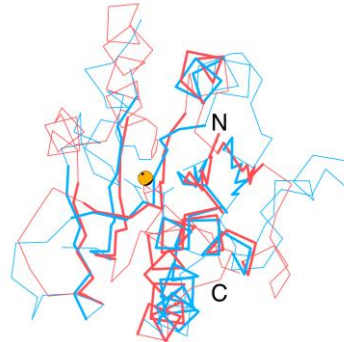
Multiple sequence alignment

BSUB00 RMAHYDSLTDLPNRHAISHLTKVLNREHSLHYNTVVFFLDLNRFKVINNDAL
ECU738 VMSTRDGM TG VYNRRHWETMLRNEFDNCRHRNRDATLLIIDIDHFKSINDTW
D90790 HEVGM DVLTKLLNRFLPTIFKREIAHANRTGTPLSVLIIDVDKFKIINDTW
SYCSLL QISSLDALTQVGNRYLFDSTLEREWQRLQRIREPLALLLC D V D F F K G F N D N Y
ECAE00 NIAHRDPLTNIFNRYNFFNEL--TVQSASAQKTPYCV MIM D I D H F K K V N D T W
AF0348 QAANVDSL TGLANRAAYNAHM-ERLTAADAPS--IGLLLI D V D R L K Q V N D I L
D90796 IRSNM D V L T G L P G R R V L D E S F D H Q L R N A E P L N -- L Y L M L L D I D R F K L V N D T Y
Y4LL_R HMARHDALTGLPNRQFLREEF-ERLSDHIAPSTR LAILCL D L D G F K A I N D A Y
Y07I_M YLADHDDL TGLHNRALLQHLDQRLAPGQPGP--VAALFL D L D R L K A I N D Y L
.....

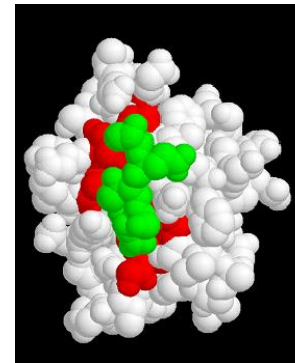
Protein similarity search and family identification/classification (conserved motif)



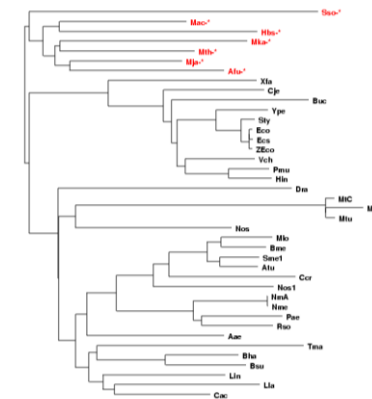
Structure modeling (Protein,RNA)



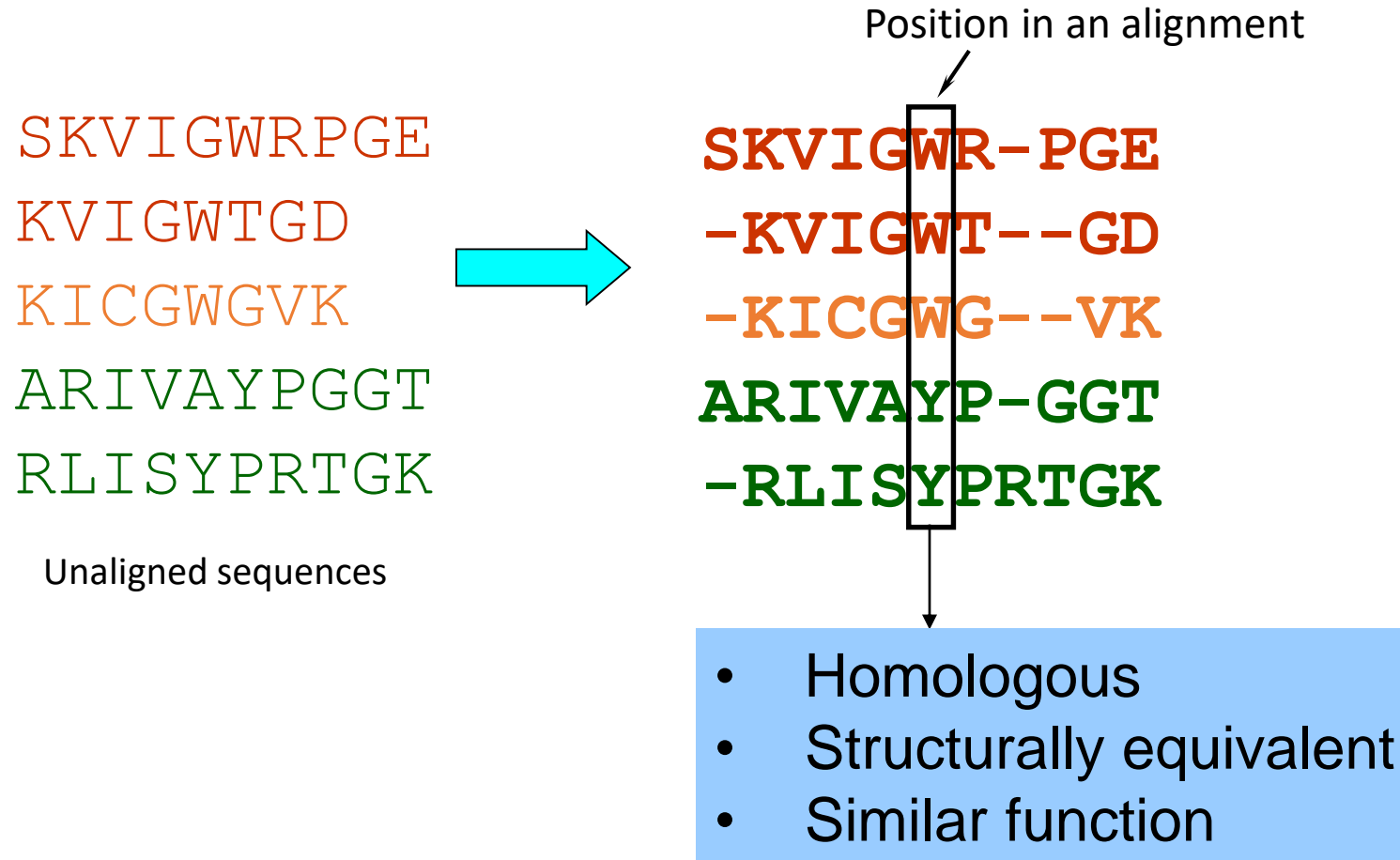
Active site prediction experimental design



Phylogenetic analysis



Meaning of alignments



Outline

- Multiple sequence alignment (MSA)
- Scoring MSA
- MSA algorithms
 - Exact approach - Dynamic programming
 - Progressive alignments - ClustalW
 - Iterative approach - MUSCLE
 - Profile-based approach - Promals
- Warning

Scoring a multiple sequence alignment

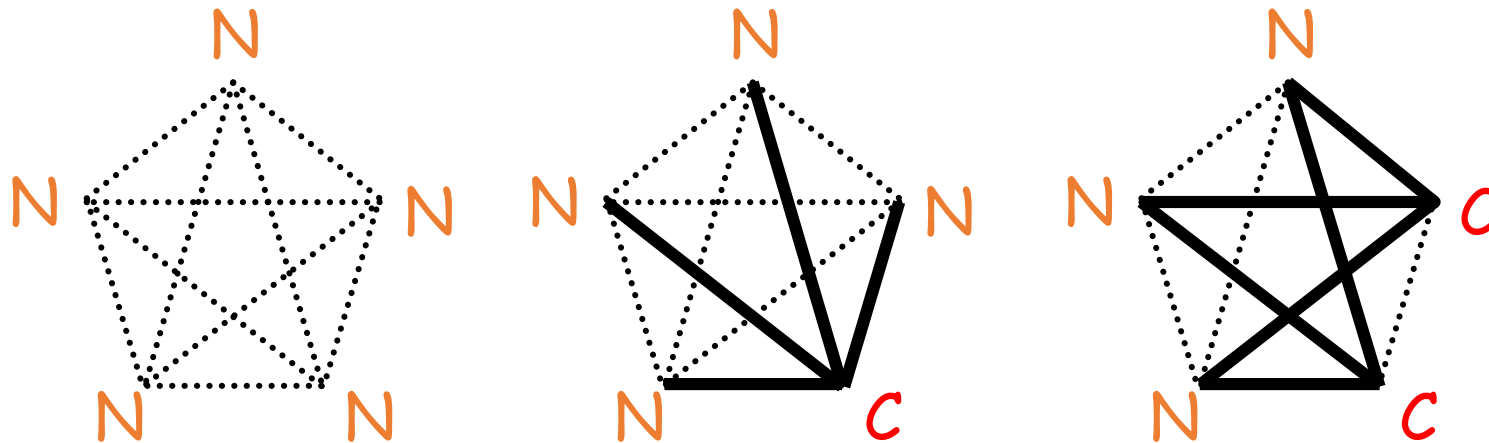
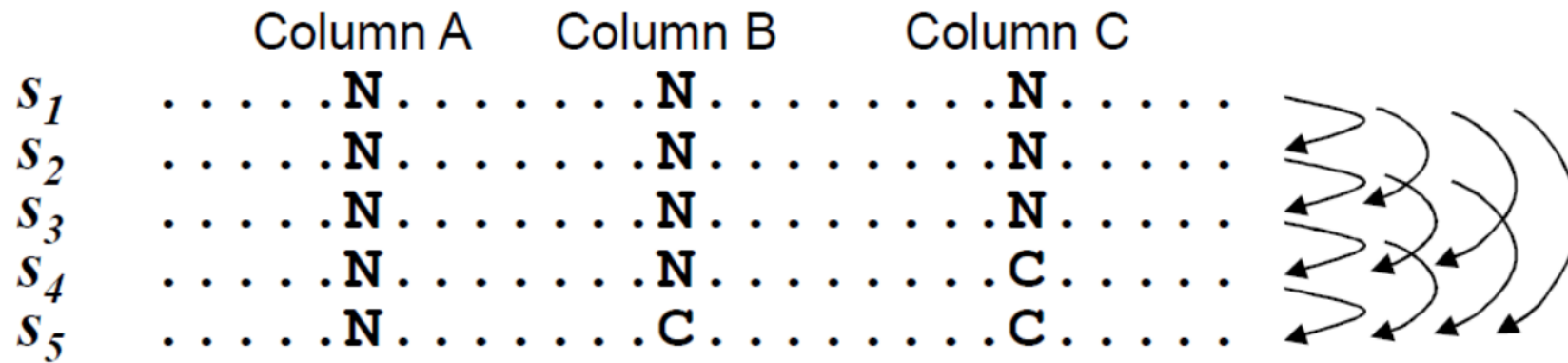
A	T	_	G	C	G	A	A	T	_	G	C	G	A	A	T	_	G	C	G	A
A	_	C	G	T	_	A	A	C	_	G	T	_	A	A	C	_	G	T	A	_
A	T	C	A	C	_	A	A	T	C	A	C	_	A	A	T	C	A	C	A	_

- Score: more conserved columns, better alignment
- To find alignment that maximizes a score function

Sum-of-pairs (SP) scoring

- Standard MSA scoring method
- Assumes column independence
- SP is a column-by-column cost/weight function
- SP scored using a **substitution matrix** (e.g PAM or BLOSUM)
- MSAs *maximize* total alignment score by *maximizing* each column SP score

Scoring a multiple sequence alignment: sum-of-pairs (SP) scoring



Alignmt	Score	Column		
		A	B	C
N – N	6	10	6	3
N – C	-3	0	4	6
C – C	9	0	0	1
		60	24	9

History of MSA

1975 Sankoff

Formulated multiple alignment problem and gave DP solution

1988 Carrillo-Lipman

Branch and Bound approach for MSA

1990 Feng-Doolittle

Progressive alignment

1994 Thompson-Higgins-Gibson-ClustalW

Most popular multiple alignment program

1998 DIALIGN *(Segment-based multiple alignment)*

2000 T-coffee *(consensus-based)*

2004 MUSCLE

2005 Kalign

2005 ProbCons *(uses Bayesian consistency)*

2006 M-Coffee *(consensus meta-approach)*

2006 Espresso *(3D-Coffee; use structural template)*

2007 PROMALS *(profile-profile alignment)*

2009 FastTree

Outline

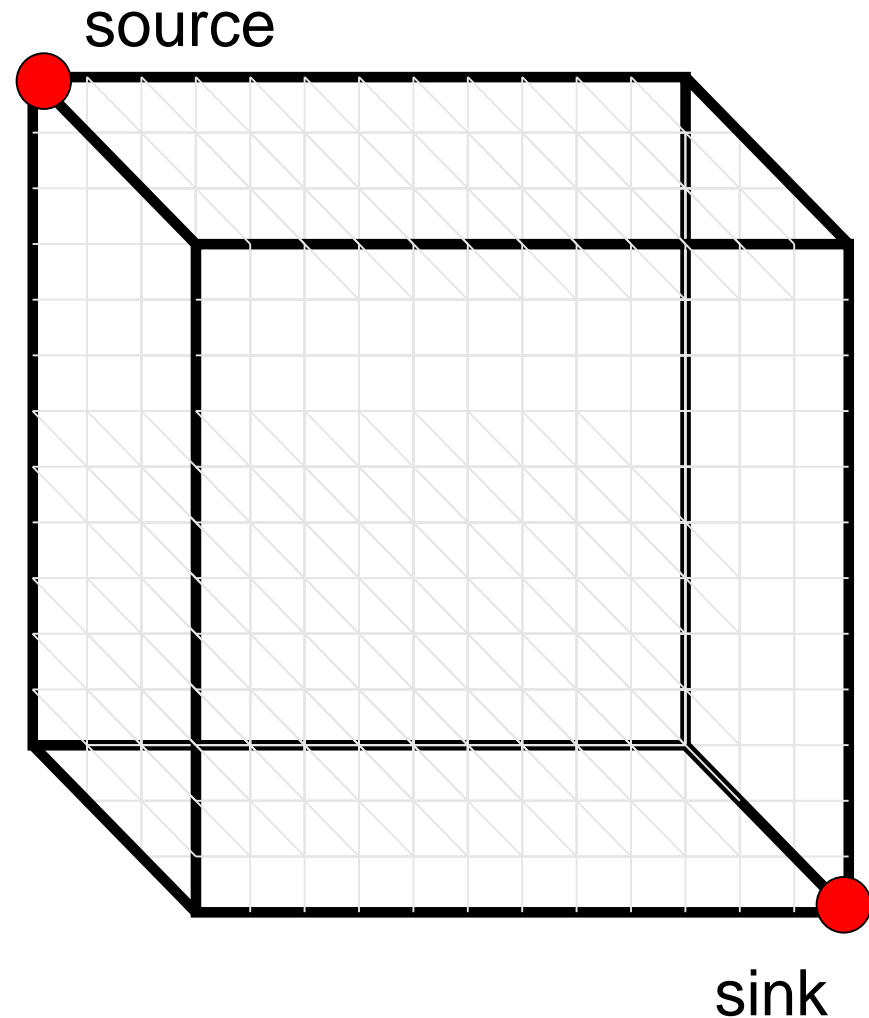
- Multiple sequence alignment (MSA)
- Scoring MSA
- MSA algorithms
 - Exact approach - Dynamic programming
 - Progressive alignments - ClustalW
 - Iterative approach - MUSCLE
 - Profile-based approach - Promals
- Recommendations

Exact approach

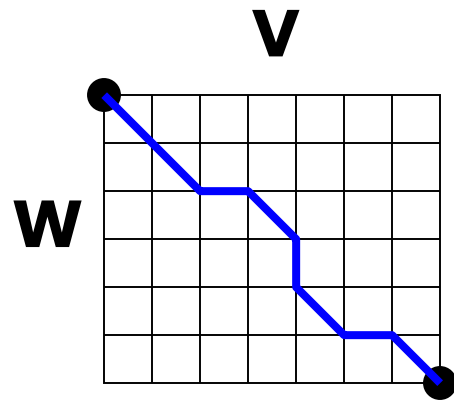
- Exact methods of multiple alignment use dynamic programming (Generalization of Needleman-Wunsch)
- Guaranteed to find optimal solutions
- Computationally expensive and so impractical
 - Time grows as product of sequence lengths

From pairwise to multiple alignment

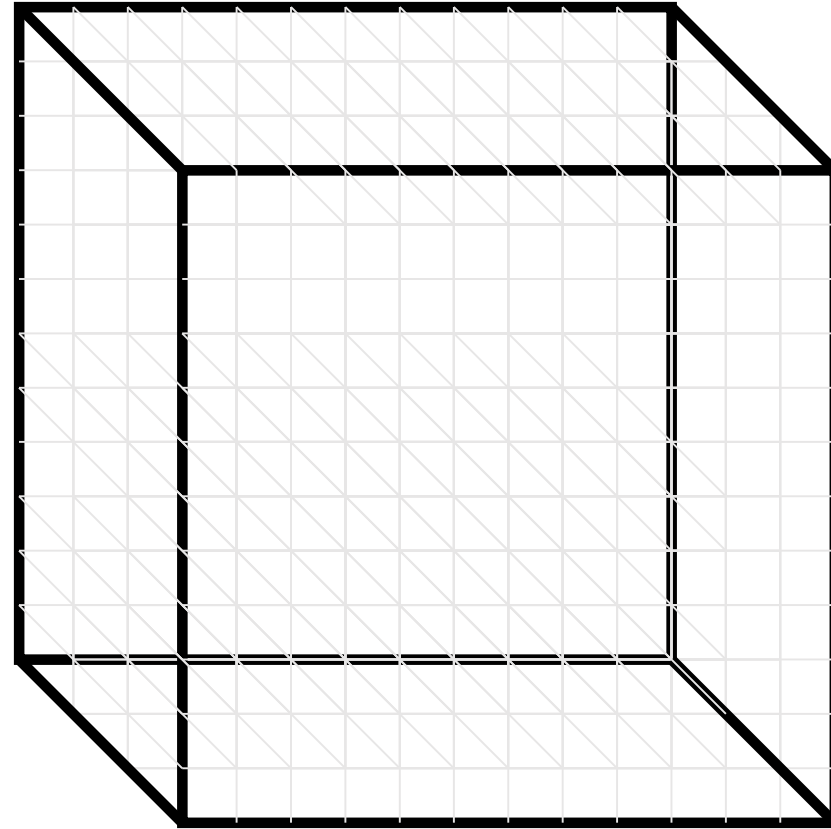
- Alignment of 2 sequences is represented as a 2-row matrix
- In a similar way, we represent alignment of 3 sequences as a 3-row matrix (or a 3-D “Manhattan Cube”, with each axis representing a sequence to align)
- For global alignments, go from source to sink



2D vs 3D alignment cells

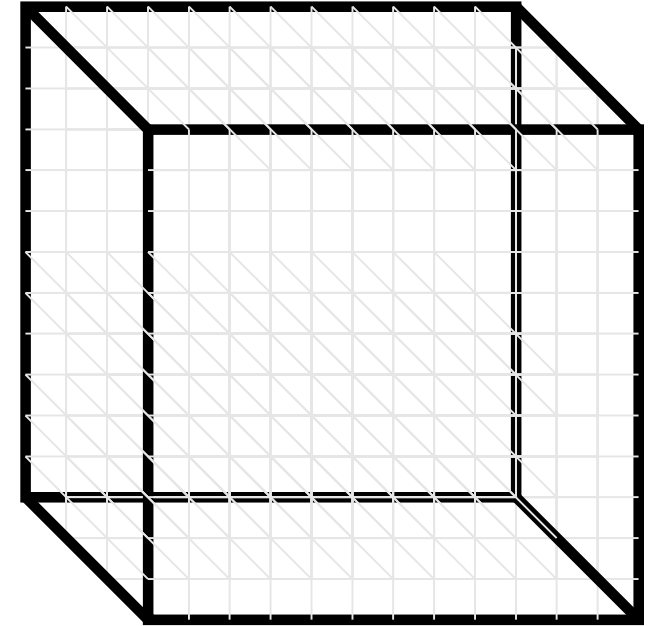


2D table



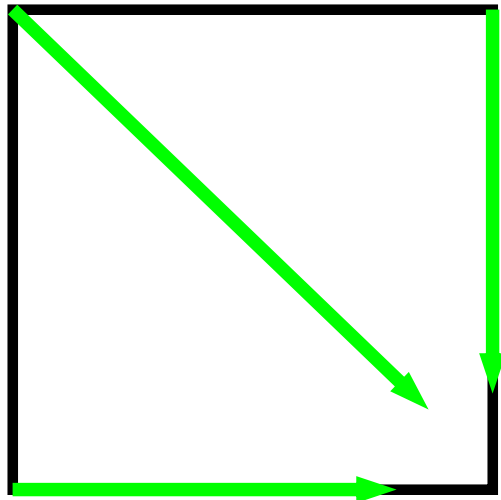
3D graph

0	1	2	2	3	4	5	6
	A	T	_	G	C	G	A
0	1	1	2	3	4	4	5
	A	_	C	G	T	_	A
0	1	2	3	4	5	5	6
	A	T	C	A	C	_	A

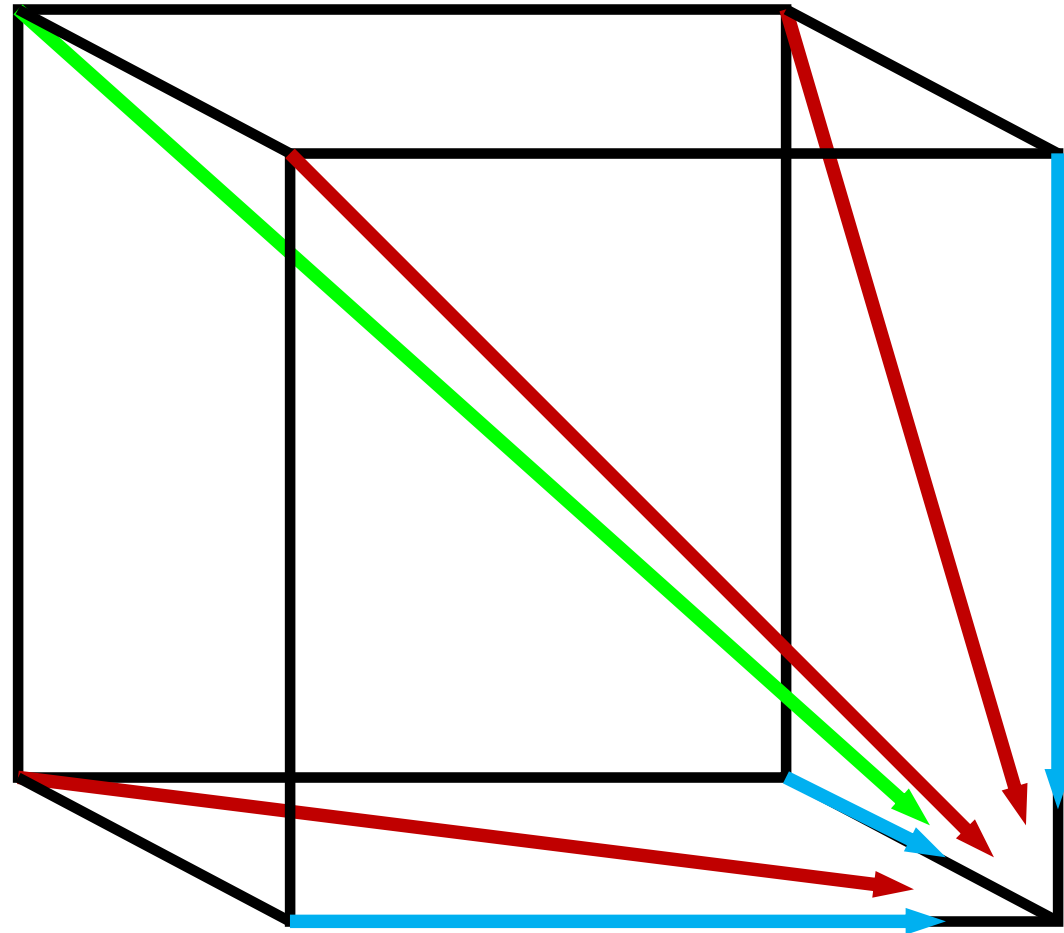


$(0,0,0) \rightarrow (1,1,1) \rightarrow (2,1,2) \rightarrow (2,2,3) \rightarrow (3,3,4) \rightarrow$
 $\rightarrow (4,4,5) \rightarrow (5,4,5) \rightarrow (6,5,6)$

2D vs 3D alignment cell: 3 paths vs 7 paths



Pairwise: 3 possible paths (match/mismatch, insertion, and deletion)



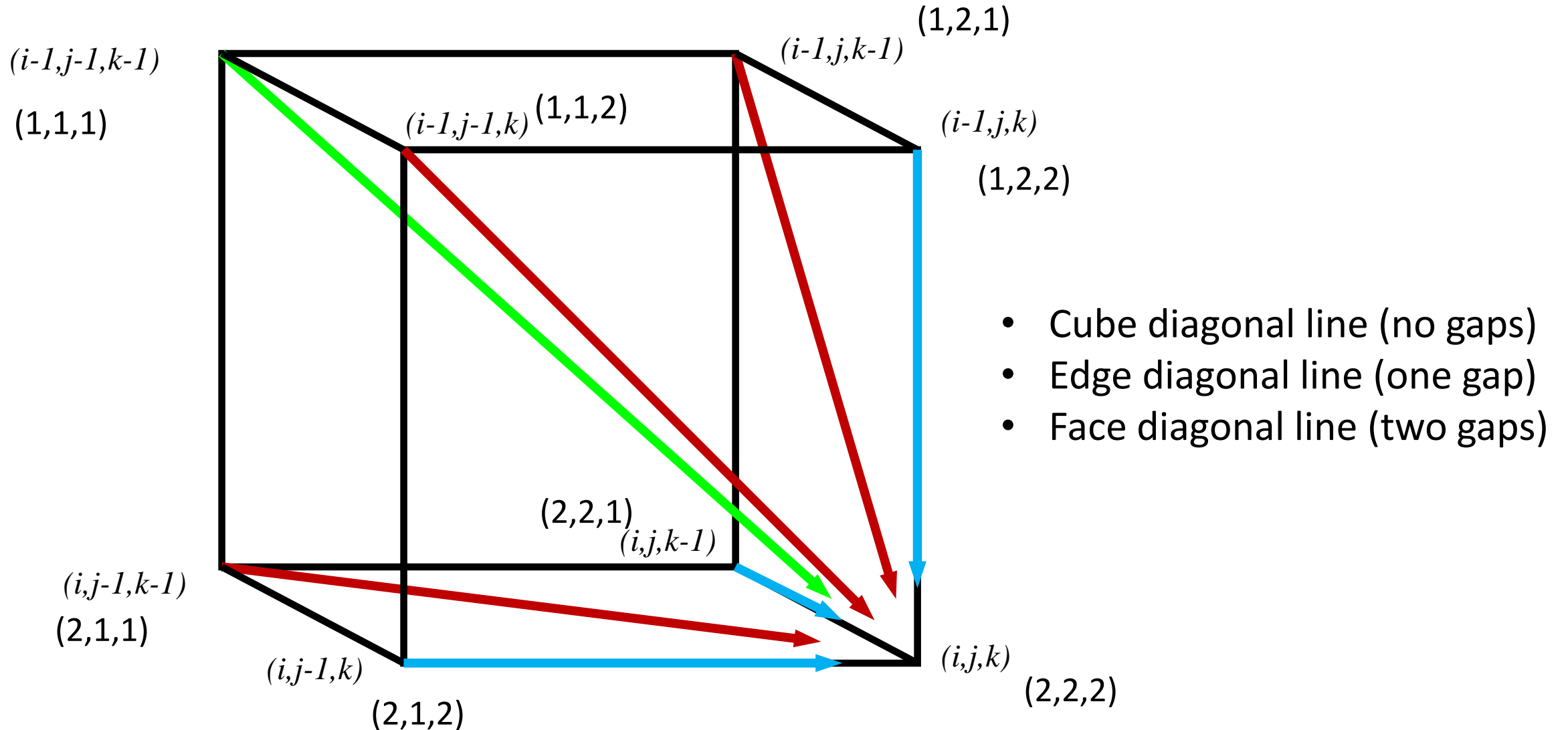
In **3-D**, 7 edges in each unit cube

There are seven cases when aligning three sequences

1	2	3	4	5	6	7
I	I	I	-	I	-	-
J	J	-	J	-	J	-
K	-	K	K	-	-	K

$2^3 - 1$ to choose the maximum similarity

Architecture of 3D alignment cell



Multiple alignment: dynamic programming

- $$s_{i,j,k} = \max \left\{ \begin{array}{ll} s_{i-1,j-1,k-1} + \delta(v_i, w_j, u_k) & \text{cube diagonal:} \\ s_{i-1,j-1,k} + \delta(v_i, w_j, _) & \text{no gaps} \\ s_{i-1,j,k-1} + \delta(v_i, _, u_k) & \\ s_{i,j-1,k-1} + \delta(_, w_j, u_k) & \text{face diagonal:} \\ s_{i-1,j,k} + \delta(v_i, _, _) & \text{one gap} \\ s_{i,j-1,k} + \delta(_, w_j, _) & \\ s_{i,j,k-1} + \delta(_, _, u_k) & \text{edge diagonal:} \\ & \text{two gaps} \end{array} \right.$$

- $\delta(x, y, z)$ is an entry in the 3D scoring matrix

MSA: running time

- For 3 sequences of length n , operation time is $7n^3$; $O(n^3)$
- For k sequences, build a k -dimensional Manhattan, with operation time $(2^k-1)(n^k)$; $O(2^k n^k)$
 - 32 thousand years for 10 seqs of 100 residues!
- Conclusion: although dynamic programming approach for alignment between two sequences is easily extended to k sequences (simultaneous approach), it is impractical due to exponential running time.
- Heuristic sequence alignment algorithm is needed, which doesn't guarantee to find the optimal solution

Outline

- Multiple sequence alignment (MSA)
- Scoring MSA
- MSA algorithms
 - Exact approach - Dynamic programming
 - Progressive alignments - ClustalW
 - Iterative approach - MUSCLE
 - Profile-based approach - Promals
- Recommendations

Progressive alignment

- Feng & Doolittle 1987, Higgins and Sharp 1988
- Concept: to build the alignment of larger number of sequences from partial alignments of subsets of sequences
- A guide tree (related to a phylogenetic tree) is used to determine how to combine pairwise alignments one by one to create a multiple alignment.
- Examples: ClustalW

ClustalW – the most widely used program

ClustalW

Valid format for input is: FASTA(Pearson)
max number of sequences = 30
max total length of sequences = 10000

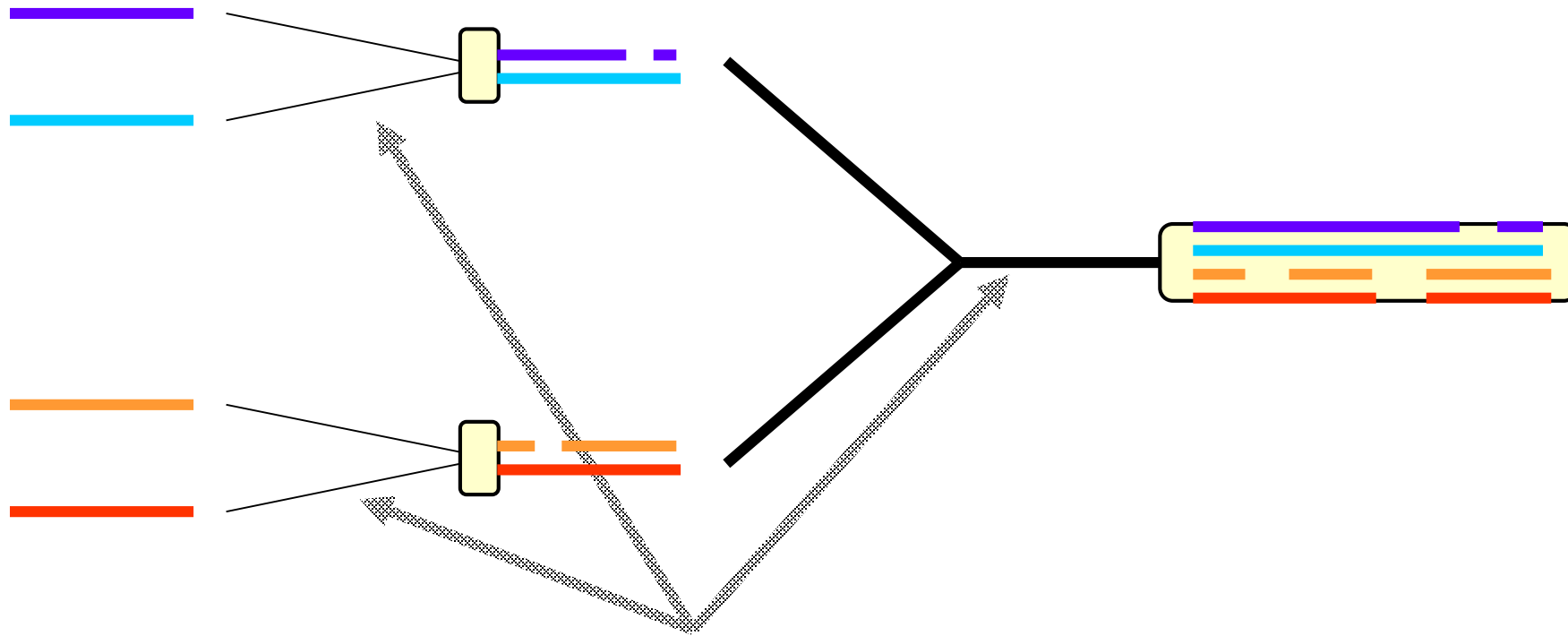
[Help page](#)

[More information on Clustal](#) [home page](#)

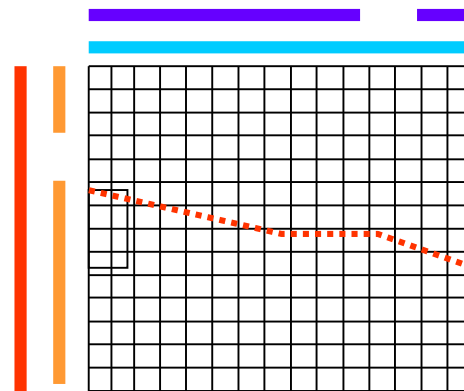
Scoring matrix :	<input type="text" value="Blosum"/>		
Opening gap penalty :	<input type="text" value="10"/>	Extending gap penalty :	<input type="text" value="0.05"/>
End gap penalty :	<input type="text" value="10"/>	Separation gap penalty :	<input type="text" value="0.05"/>
Output format :	<input type="text" value="Clustal"/>	Output order :	<input type="text" value="Input"/>

Input sequences:
(see above for valid
formats)

ClustalW algorithm



Dynamic Programming Using A Substitution Matrix



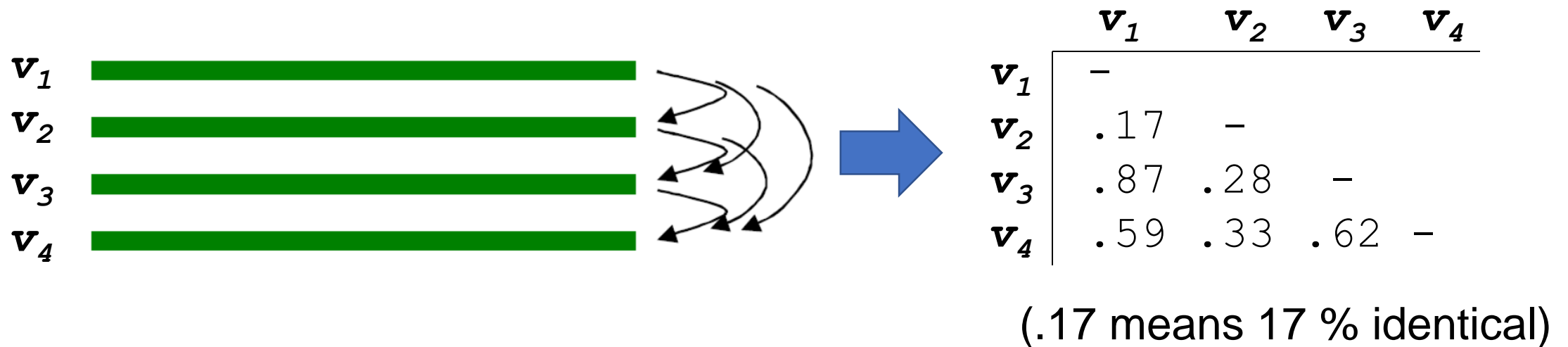
ClustalW

The three basic steps in the CLUSTAL W approach are shared by all progressive alignment algorithms:

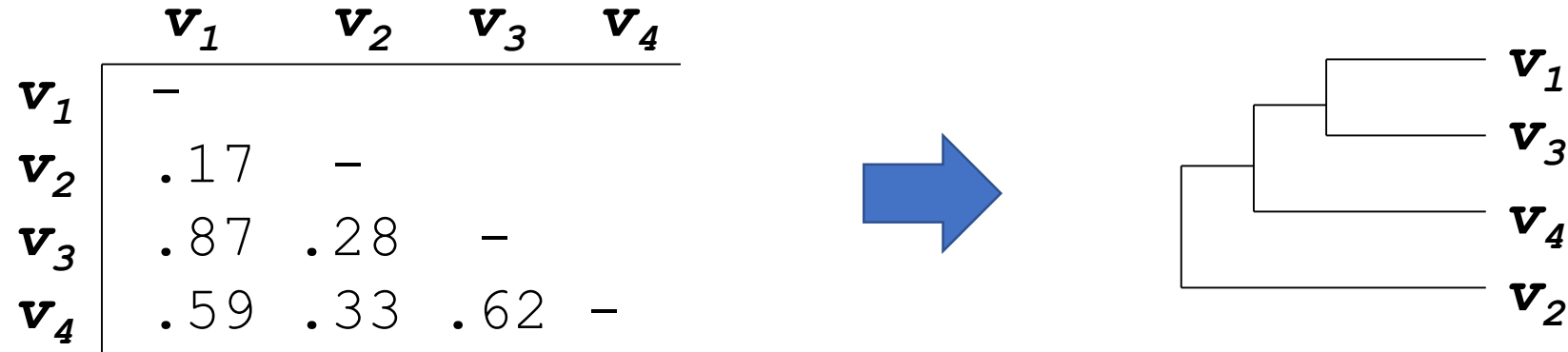
- A. Calculate a matrix of pairwise distances based on pairwise alignments between the sequences
- B. Use the result of A to build a guide tree, which is an inferred phylogeny for the sequences
- C. Use the tree from B to guide the progressive alignment of the sequences

Step 1: Pairwise alignment

- Aligns each sequence against each other using dynamic programming
- a similarity or distance measure for the pair is calculated using the aligned portion (gaps excluded) - for example, percent identity.
- Similarity = exact matches / sequence length (percent identity)

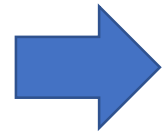
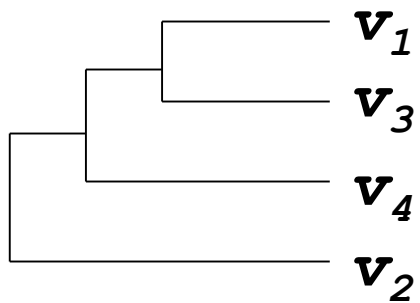


Step 2: Guide tree by clustering



- To build guide tree
 - Neighbour-Joining (NJ)
 - Unweighted pair group method using arithmetic averages (UPGMA)
- Guide tree roughly reflects evolutionary relations

Step 3: progressive alignment of the sequences




calculate:

$$\begin{aligned} V_{1,3} &= \text{alignment}(v_1, v_3) \\ V_{1,3,4} &= \text{alignment}((V_{1,3}), v_4) \\ V_{1,2,3,4} &= \text{alignment}((V_{1,3,4}), v_2) \end{aligned}$$

- Partial alignment was generated (profile)
- In the past we were aligning a **sequence against a sequence**
- Can we align a **sequence against a profile**?

Scoring an alignment of two partial alignments

1	peeksav	t	a	l	
2	geekaav	l	a	l	
3	padktnv	k	a	a	
4	aadktnv	k	a	a	
					
5	egewql	v	l	h	v
6	aaektki	r	s	a	

Sequence weights

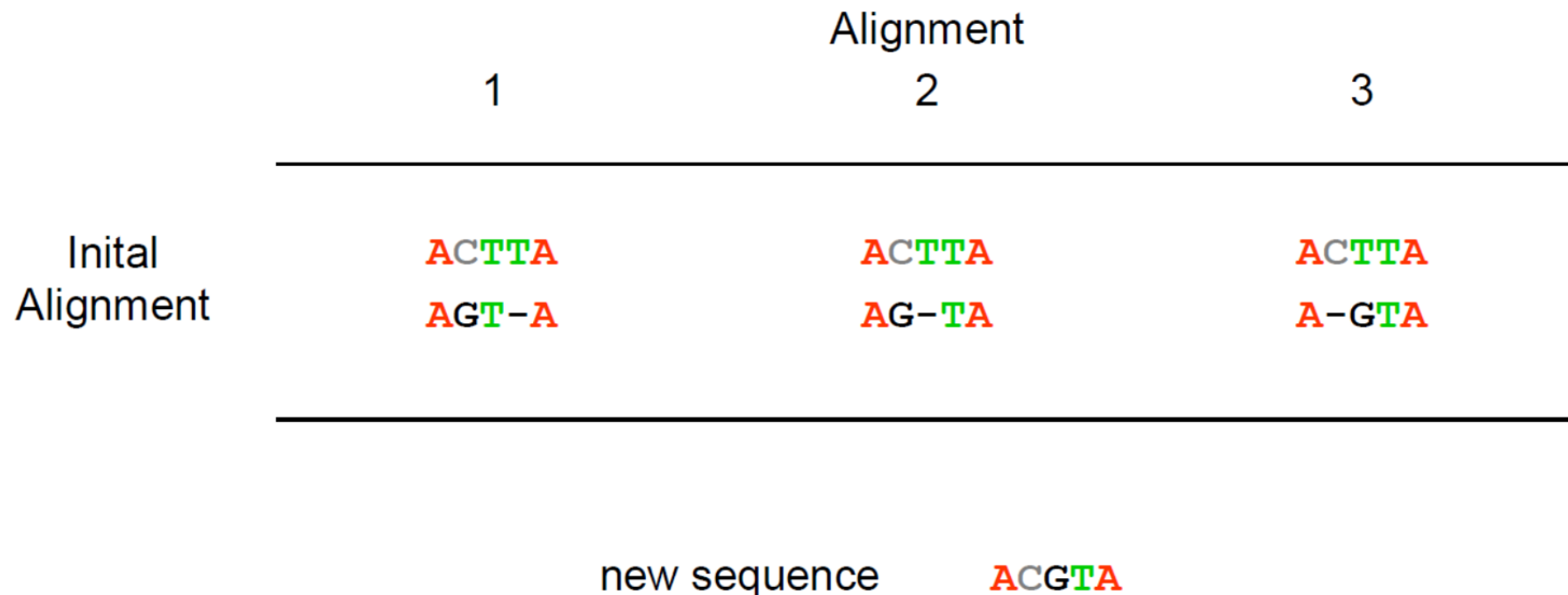
w_1, \dots, w_6

‘W’ stands for ‘weighted’
(sequences are weighted
differently).

Score: $\frac{1}{8} [M(t, v)w_1w_5 + M(t, i)w_1w_6 + \dots + M(k, i)w_4w_6]$

Potential problems with ClustalW

- ClustalW is a “greedy” algorithm
 - makes the best immediate solution (local choice) in hopes of finding the best overall (global) solution
 - choices are made regardless of later consequences
 - early mistakes get propagated throughout the rest of the alignment





Potential problems with ClustalW

- ClustalW is a “greedy” algorithm
 - makes the best immediate solution (local choice) in hopes of finding the best overall (global) solution
 - choices are made regardless of later consequences
 - early mistakes get propagated throughout the rest of the alignment

	Alignment		
	1	2	3
Initial Alignment	ACTTA AGT-A	ACTTA AG-TA	ACTTA A-GTA
Later Alignment	ACTTA AGT-A ACGTA	ACTTA AG-TA ACGTA	ACTTA A-GTA ACGTA


Clustal Omega

- Profile HMMs to model groups of sequences whereas Clustal W uses sequence profiles to store information about groups of sequences




Clustal: Multiple Sequence Alignment

Multiple alignment of nucleic acid and protein sequences



Clustal Omega

- Latest version of Clustal - fast and scalable (can align hundreds of thousands of sequences in hours), greater accuracy due to new HMM alignment engine
- Command line/web server only (GUI public beta available soon)



ClustalW/ClustalX

- "Classic Clustal"
- GUI (ClustalX), command line (ClustalW), web server versions available

<http://www.clustal.org/>

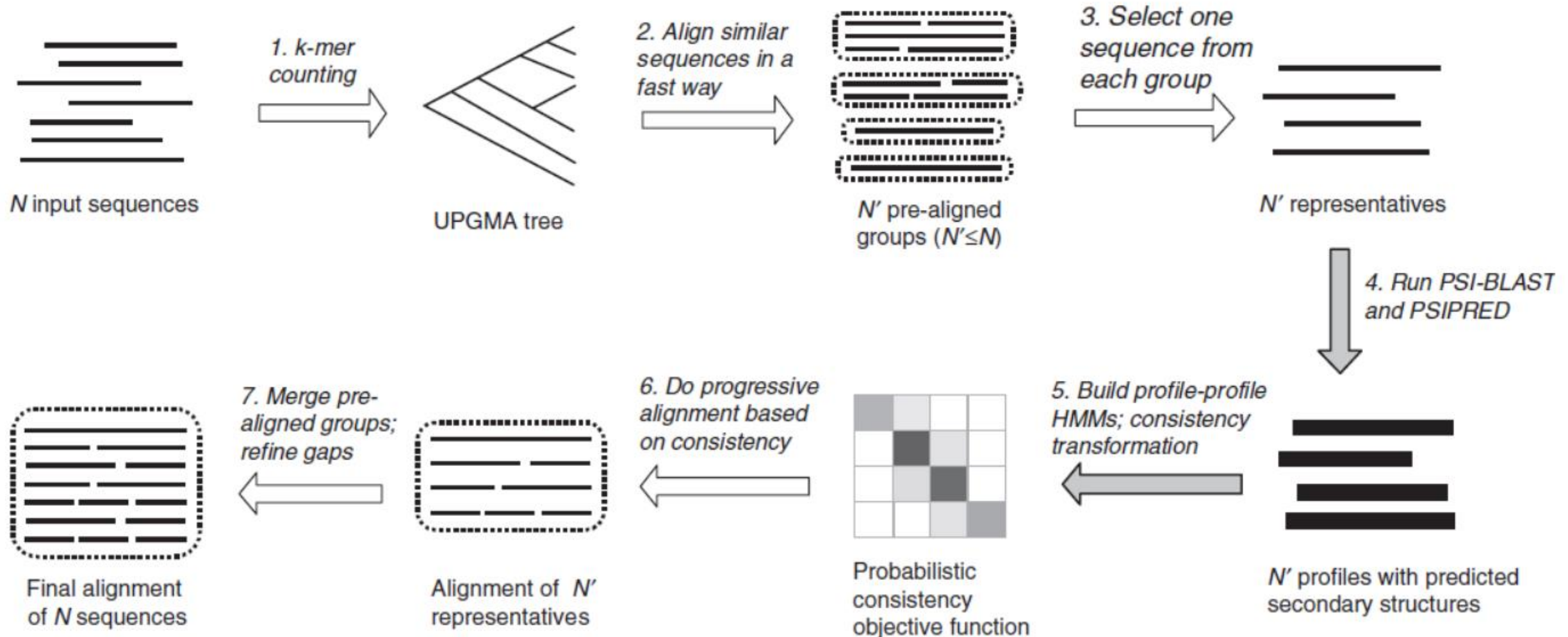
Iterative alignment

- Progressive alignment:
 - The order of selection of sequences can influence the alignment
 - Once there is a gap, always a gap
- How to avoid committing to a non-optimal pairwise decision?
 - Revisit alignments
 - This is the focus of iterative alignments

Iterative alignment

- Basic iterative refinement algorithm
 - Remove a sequence from the current multiple alignment
 - Realign the removed sequence back to the multiple alignment
 - Repeat until removal and realignment of any sequence does not improve the alignment score
- MUSCLE (multiple sequence alignment by log-expectation)

Profile-HMM method: Promals



http://prodata.swmed.edu/promals

The PROMALS multiple sequence alignment server

PROMALS constructs multiple protein sequence alignments using information from database searches and secondary structure prediction. [\[Documentation\]](#)

Enter your sequences in [FASTA](#) format:

Or upload a local file containing your sequences:

Enter your [email](#) address to receive the result ([recommended](#)):

Alignment options:

- [Weight for amino acid scores](#):
- [Weight for predicted secondary structure scores](#):
- [Identity threshold above which fast alignment is applied](#):

Enter a name for your job ([recommended](#)):

[PROMALS Documentation](#)

[Reference](#): Pei, J. and Grishin, N. V. (submitted). Towards accurate multiple sequence alignments of distantly related proteins.

Comments, suggestions and bug reports to: jpei@chop.swmed.edu

History of MSA

1975 Sankoff

Formulated multiple alignment problem and gave DP solution

1988 Carrillo-Lipman

Branch and Bound approach for MSA

1990 Feng-Doolittle

Progressive alignment

1994 Thompson-Higgins-Gibson-ClustalW

Most popular multiple alignment program

1998 DIALIGN (Segment-based multiple alignment)

2000 T-coffee (consensus-based)

Acceptable result

2004 MUSCLE

2005 Kalign

Fast and acceptable result, gappy

2005 ProbCons (uses Bayesian consistency)

2006 M-Coffee (consensus meta-approach)

2006 Espresso (3D-Coffee; use structural template)

2007 PROMALS (profile-profile alignment)

Slow but most accurate

2009 FastTree

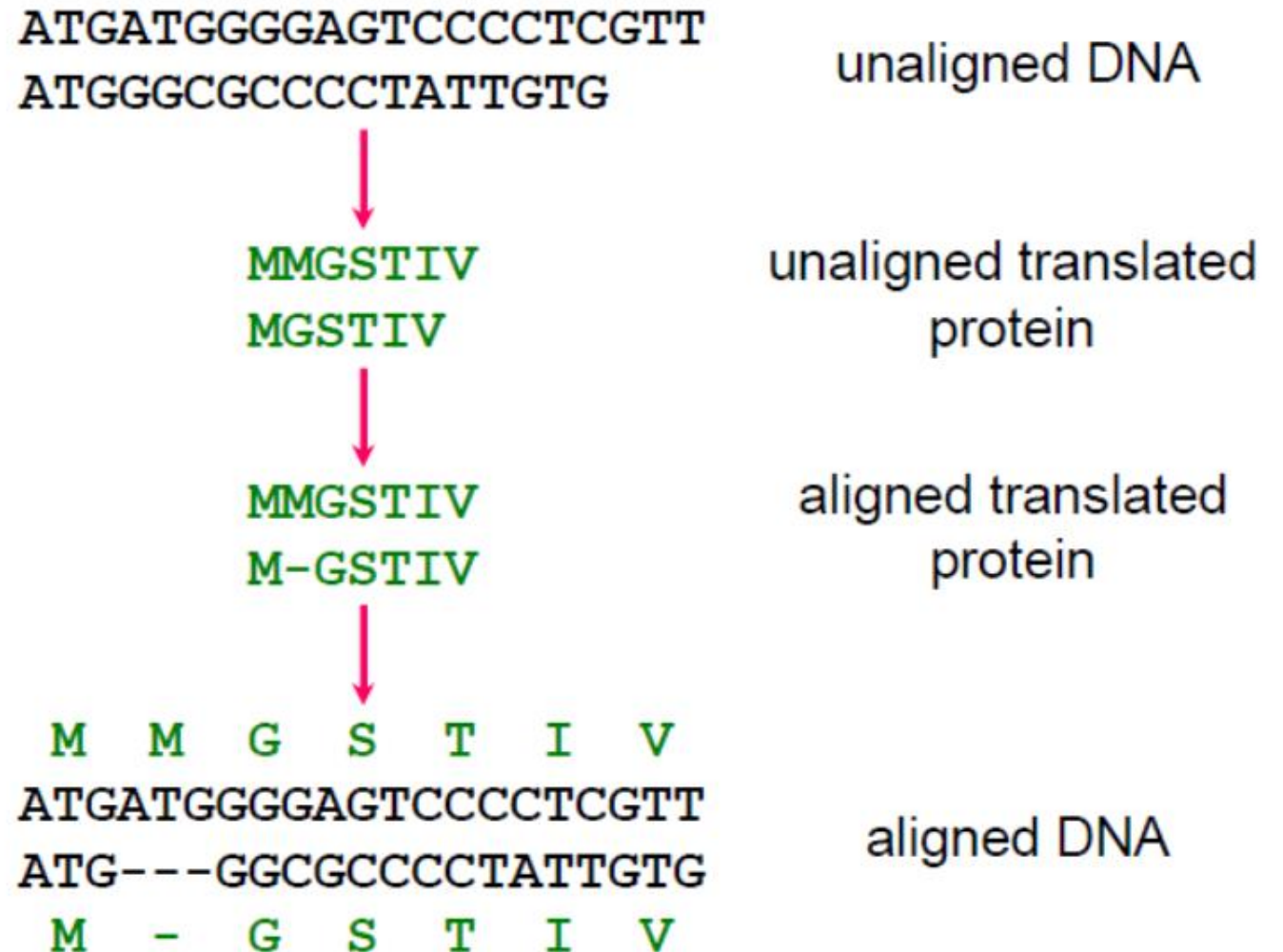
Fast and working with large datasets; ok

Recommendations

- Many dozens of MSA programs have been introduced in recent years. None is optimal. Each offers unique strengths and weaknesses.
- MSA algorithms assume that sequences are homologous
 - MSA programs will align anything and all sequences, even if they are not homologous.
- Ideally sequences with one domain or sequences with same domain architecture
- Proteins are easier to align than DNA
- If it looks wrong it probably is wrong!
- Manual alignment is needed

Proteins are easier to align than DNA

- Therefore, if your DNA sequences are too divergent try aligning their amino acid translation, and then translating the sequence back to DNA



Multiple sequence alignment editors

- BioEdit - MS-Windows
- Genedoc - MS-Windows
- EditSeq/MegAlign - Lasergene - Mac or MS-Windows
- DNA Strider - Macintosh
- Seq-AL - Macintosh
- ASAD - Excel - Macintosh or MS-Windows
- SeqPup - Mac. MS-Windows, X-Windows

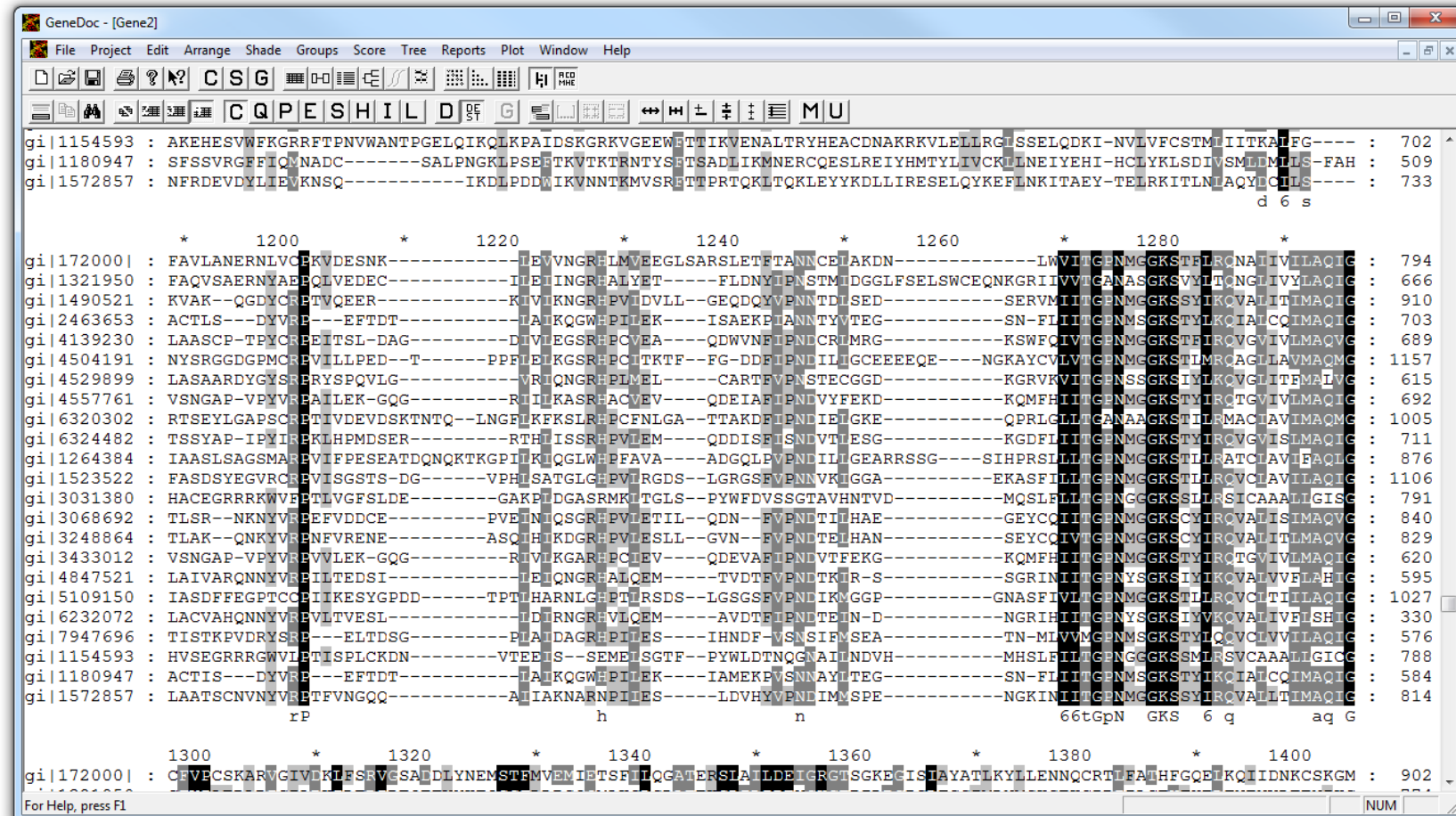
MSA-Visualization and improvement

- GeneDoc (Windows)
- Download: <http://genedoc.software.informer.com/download/>
 - Arranging and Editing
 - GeneDoc's Grab and Drag arrangement mode allows you to move residues around like beads on a string
 - Shading Alignments
 - Reports: Stats, Score, Composition
 - Exporting and Copying Figures

<http://www.nrbsc.org/old/gfx/genedoc/gdpaf.htm>

GeneDoc: Conservation Mode

- GeneDoc (Windows)



GeneDoc: Property Mode



ClustalW

CLUSTAL W (1.83) multiple sequence alignment

```

beta globin  -----MVHLTPEEKSAVTALWGKVNVDD--EVGGEALGRLLVVYPWTQRFFESFG- 47
myoglobin   -----MGLSDGEWQLVLNVWVGKVEADIPGHGQEVLIQLFKGHPEETLEKFDKFK- 48
neuroglobin -----MERPEPELIRQSWRAVSRSPLEHGTVLFAFLFALEPDLLPLFQYNCR 47
soybean      -----MVAFTPEKQDALVSSSFQAFKANIPQYSVVFYTSILEKAPAAKDLFSFLA- 49
rice         MALVEDNNAVAVSFSEEQEALVLKSWAILKKDSANIALRFFLKIFEVAPSASQMFSFLR- 59
              :   :   :   :   .   .   .   :   :   *   *   .
              ▽
beta globin  DLSTPDAVMGNPKVKAHGKKVLGAFSDGLAHLDNLKGTFATLS-----ELHCDKLHVDPE 102
myoglobin   HLKSEDEMKALEDLKKHGATVLTALGGILKKKGHHAEIKPLA-----QSHATKHKIPVK 103
neuroglobin QFSSPEDCLSSPEFLDHIRKVMLVIDAAVTNVEDLSSLEEYLA---LGRKHRAVGVKLS 104
soybean      --NGVDPT--NPKLTGHA EKLFALVRDSAGQLKASGTVVADAA---LGSVHAQKAVTDP 101
rice         --NSDVPLEKNPKLKTHAMSVFVMTCEAAQAQLRKAGKVTVRDTTLKRLGATHLKYGVGDA 117
              .   .   .   *   .::   :   :   :
              :   :   :   :   :   :   :   :   :
beta globin  NFRLLGNVLVLCVLAHGF-GKEFTPPVQAAYQKVVAGVANALAHKYH----- 147
myoglobin   YLEFISECIIQVLQSKH-PGDFGADAQGAMNKALELFRKDMASNYKELGFQG 154
neuroglobin SFSTVGESLLYMLEKCL-GPAFTPATRAAWSQLYGAVVQAMSRGWDGE---- 151
soybean      QFVVVKEALLKTIKAAV-GDKWSDLSRAWEVAYDELA AAIKKA----- 144
rice         HFEVVKFALLDTIKEEVPADMWS PAMKSAWSEAYDHLVAAIKQEMKPAE--- 166
              :   :   ::   :   :   *   .   .   :

```

Note how the region of a conserved histidine (▼) varies depending on which of five prominent algorithms is used

Praline

(a) Praline multiple sequence alignment

```

beta globin      .....MVHLTPEEKSAVTALWGKV..NVDEVGGEALGRLLVVYPWTQRFFES.FG
myoglobin       .....MGLSDGEWQLVLNVWGKVEADI PGHGQEV LIRLFKGH PETLEKFDK.FK
neuroglobin     .....MERPEPELI RQSWRAVSRSPLEHGT VLFARLFALEPDLLPLFQYNCR
soybean         .....MVAFT EKQDALVSSSF EAFKANIPQYSVVFYTSILEKAPAAKD LFS..FL
rice            MALVEDNNAVAVSFSEEQEALVLKSWAILKKDSANIALRFFLKIFEVAPSASQMFS..FL
Consistency     000000000014265438257934573463364343624453686433*35344*50063

beta globin      DLSTPDAVMGNPKVKAHGKKVLGAFSDG LAHLDNLKGT FATLSEL..HCDKLH....VDP
myoglobin       HLKSEDEMKASEDLKKHGATVLTALGGILKKKGHHEAEIKPLAQS..HATKHK....IPV
neuroglobin     QFSSPEDCLSSPEFLDHIRKVMLVIDAAVTNVEDLSSLEEYLASLGRKHRAVG....VKL
soybean         A.NGVDP..TNPKLTGHA EKLFALVRDSAGQL.KASGT VVADAA....LGSVHAQKAVTD
rice            R.NSDVPLEKNPKLKTHAMSVFVMTCEAAAQL.RKAGKVTVRDTTLKRLGATHLKYGVGD
Consistency     3166354224776653*4368635424454451335634333542003335440000922

beta globin      ENFRLLGNVLVCVLAHHF.GKEFTPPVQAAYQKV VAGVANALAHKYH.....
myoglobin       KYLEFISECIIQVLQSKH.PGDFGADAQGAMNKALELFRKDMASNYKELGFQG
neuroglobin     SSFSTVGESLLYMLEKCL.GPAFTPATRAAWSQLYGAVVQAMSRGWD..GE..
soybean         PQFVVVKEALLKTIKAAV.GDKWSELSRAWEVAYDELAAAIKKA.....
rice            AHFEVVKFALLDTIKEEVPADMWSPAMKSAWSEAYDHLVAAIKQEMKPAE...
Consistency     43744844498258542305336554454*55465426446754322001000
```

Note also the changing pattern of gaps within the boxed region in these five different alignments.

MUSCLE

(b) MUSCLE (3.6) multiple sequence alignment

```

beta globin  -----MVHLTPEEKSAVTALWGKVND--EVGGEALGRLLVVYPWTQRFFES-FG
myoglobin   -----MGLSDGEWQLVLNVWVKVEADIPGHGQEVLIIRLFKGH PETLEKFDK-FK
neuroglobin -----MERPEPELIHQSWRAVSRSPLEHGTVLFARLFALPDLLPLFQYNCR
soybean      -----MVAFTKQDALVSSSF EAFKANIPQYSVVFYTSILEKAPAAKDLFSF-LA
rice         MALVEDNNAVAVSFSEEQEALVLKSWAILKKDSANIALRFFLKIFEVAPSASQMFSF-LR
               :   :   :   :   . . .   .   : :   *   * .

               ▽                               ▴
beta globin  DLSTPDAVMGNPKVKAHGKKVLGAF---SDGLAHLN LKGT FATLSELHCDKLH--VDPE
myoglobin   HLKSEDEMKA SEDLKKHGATVLTAL---GGILKKKGHHEAEIKPLAQSHATKHK--IPVK
neuroglobin QFSSPEDCLSSPEFLDHIRKVMLVI---DAAVTNVEDLSSLEEYLASLGRKHRAVGKLS
soybean      NGVDP----TNPKLTGHA EKLFALVRDSAGQLKASGTVVAD----AALGSVHAQKAVTDP
rice         NSDVP--LEKNPKLKT HAMS VFVMTCEAAAQLRKAGKVTVRDTTLKRLGATHLKYGVGDA
               . . . * . : :           :           :           :

beta globin  NFRLLGNVLVLCVLAH HFGKE-FTPPVQAAYQKV VAGVANALAHKYH-----
myoglobin   YLEFISECIIQVLQSKHPGD-FGADAQGAMNKALELFRKDMASNYKELGFQG
neuroglobin SFSTVGESLLYMLEKCLGPA-FTPATRAAWSQLYGAVVQAMSRGWDGE----
soybean      QFVVVKEALLKTIKAAVGDK-WSDELSRAW EVAYDELA AAIKKA-----
rice         HFEVVKFALLDTIKEEVPADMWS PAMKSAWSEAYDHLVAAIKQEMKPAE---
               :   :   : :   :           :   *   .   .   :
  
```


Probcons

(c)
PROBCONS

beta globin	M-----VHLTPEEKSAVTALWGKVNVD--EVGGEALGRLLVVYPWTQRFFES-FG	
myoglobin	M-----GLSDGEWQLVLNVWGKVEADIPGHGQEVLIIRLFKGH	PETLEKFDK-FK
neuroglobin	M-----ERPEPELIRQSWRAVSRS	PLEHGTVLFARLFAL
soybean	M-----VAFTEKQDALVSSSF	EAFKANIPQYSVVFYTSILEKAPAAKDLFSF-LA
rice	MALVEDNNAVAVSFS	EEQEALVLKSWAILKKDSANIALRFFLKI
	* : : : : : : *	*
beta globin	DLSTPDAVMGNPKVKAHGKKVLGAFSDGLAHLD---NLK---GTFATLSELHCDKLH	VDP
myoglobin	HLKSEDEMKA	SEDLKKHGATVLTALGGI---
neuroglobin	QFSSPEDCLSSPEFLDH	IRKVMLVIDAAVTNVEDLSSLE---
soybean	NGVDP----	TNPKLTGHA
rice	NSDVP--LEKNPKLKT	HAMSVFVMTCEAAAQLRKAGKVTVRDTTLKRLGATHLKY-GVGD
	. : . . . * : : . * *	:
beta globin	ENFRLLGNVLVCVLAHHF-GKEFTPPVQAAYQKVVAGVANALAHK-----YH	
myoglobin	KYLEFISECIIQVLQSKH-PGDFGADAQGAMNKALELFRKDMASNYKELGFQG	
neuroglobin	SSFSTVGESLLYMLEKCL-GPAFTPATRAAWSQLYGAVVQAMSRG---W-DGE	
soybean	PQFVVVKEALLKTIKAAV-GDKWSELSRAWEVAYDELA	AAIK-----KA
rice	AHFEVVKFALLDTIKEEVPADMWS	PAMKSAWSEAYDHLVAAIKQE---MKPAE
	:	:

TCoffee

(d)

CLUSTAL FORMAT for T-COFFEE Version_5.13

```

beta globin  -----MVHLTPEEKSAVTALWGKVNVDD--EVGGEALGRLLVVYPWTQRFFESFG
myoglobin   -----MGLSDGEWQLVLNVWGKVEADIPGHGQEVLIIRLFKGHPEKLEKFD-KFK
neuroglobin -----MERPEPELIQSWRAVSRSPLEHGTVLFARLFALPDLLPLFQYNCR
soybean     -----MVAFTTEKQDALVSSSFEEAFKANIPQYSVVFYTSILEKAPAAKDLFS-FLA
rice        MALVEDNNAVAVSFS EEQEALVLKSWAILKKDSANIALRFFLKIFEVAPSASQMFS-FLR
              :   :   :   :   . . .   .   : :   *   * .
              |   |   |   |   |   |   |   |   |   |   |
beta globin  DLSTPDAVMGNPKVKAHGKKVLGAFSDGLAHLDNL---KGTF---ATLSELHCDKLHVDP
myoglobin   HLKSEDEMKA SEDLKKHGATVLTAL---GGILKKKGHHEAE---IKPLAQSHATKHKIEV
neuroglobin QFSSPEDCLSSPEFLDHIRKVMLVIDAAVTNVEDL---SSLEEYLA SLGRKH-RAVGVKL
soybean     NGVDP----TNPKLTGHA EKL FALVRDSAGQLKASGTVVAD----AALGSVHAQKAVTDIP
rice        NSDVP--LEKNPKLKTHAMSVFVMTCEAA AQLRKAGKVTVRD TTKRLGATHLKYGVGDA
              .   . . . * . : :   :   :   :   :   :   :   :   :
              |   |   |   |   |   |   |   |   |   |   |   |
beta globin  ENFRLLGNVLVCVLAHHF-GKEFTPPVQAAYQKVVAGVANALAHKYH-----
myoglobin   KYLEFISECIIQVLQSKH-PGDFGADAQGAMNKALELFRKDMASNYKELGFQG
neuroglobin SSFSTVGESLLYMLEKCL-GPAFTPATRAAWSQLYGAVVQAMSRGWDG----E
soybean     Q-FVVVKEALLKTIKAAV-GDKWSD ELSRAWEVAYDELA AAIKKA-----
rice        H-FEVVKFALLDTIKEEV PADMWS PAMKSAWSEAYDHLVAAIKQE---MKPAE
              :   :   : :   :   :   :   :   :   :   :   :

```