



# ALIGNMENT FOUNDATIONS

- Why...
  - Why compare biological sequences?
- · What...
  - Alignment view of sequence changes during evolution (matches, mismatches and gaps)
- How...
  - Dot matrices
  - Dynamic programing
    - Global alignment
    - Local alignment
  - BLAST heuristic approach

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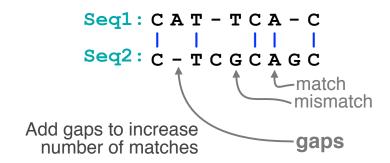
**Basic Idea**: Display one sequence above another with spaces (termed **gaps**) inserted in both to reveal **similarity** of nucleotides or amino acids.

Seq1: CATTCAC

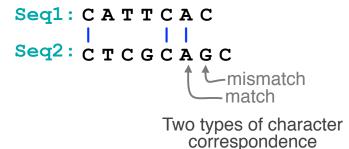
Seq2: CTCGCAGC

[Screencast Materia

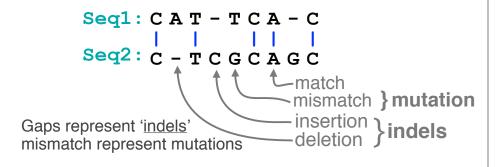
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# Why compare biological sequences?

- To obtain functional or mechanistic insight about a sequence by inference from another potentially better characterized sequence
- To find whether two (or more) genes or proteins are evolutionarily related
- To find structurally or functionally similar regions within sequences (e.g. catalytic sites, binding sites for other molecules, etc.)
- Many practical bioinformatics applications...

# Practical applications include...

- Similarity searching of databases
  - Protein structure prediction, annotation, etc...
- Assembly of sequence reads into a longer construct such as a genomic sequence
- Mapping sequencing reads to a known genome
  - "Resequencing", looking for differences from reference genome - SNPs, indels (insertions or deletions)
  - Mapping transcription factor binding sites via ChIP-Seg (chromatin immuno-precipitation sequencing)
  - Pretty much all next-gen sequencing data analysis

- applications include...

  similarity searching of databases

   Protein structure prediction

   Assembly of sequence alignment is arguably the construct such most fundamental operation of bioinformatics! ranscription factor binding sites via ChIP-Seg
  - chromatin immuno-precipitation sequencing)
  - Pretty much all next-gen sequencing data analysis

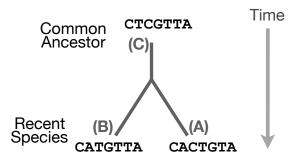
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# Sequence changes during evolution

There are three major types of sequence change that can occur during evolution.

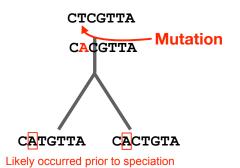
- Mutations/Substitutions
- Deletions
- Insertions



# Mutations, deletions and insertions

There are three major types of sequence change that can occur during evolution.

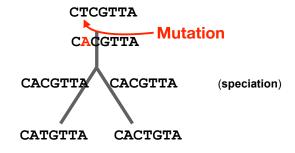
- Mutations/Substitutions CTCGTTA → CACGTTA
- Deletions
- Insertions



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- Mutations/Substitutions CTCGTTA → CACGTTA
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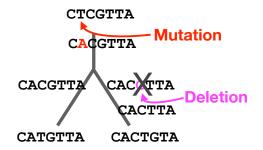
# Mutations, deletions and insertions

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- Mutations/Substitutions CTCGTTA → CACGTTA
- Deletions CACGTTA → CACGTTA

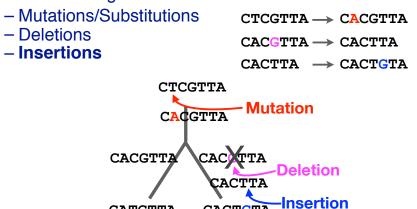
Insertions

 $CACGTTA \longrightarrow CACTTA$ 



# Mutations, deletions and insertions

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### Mutations, deletions and insertions

There are three major types of sequence change that can occur during evolution.

- Mutations/Substitutions CTCGTTA → CACGTTA
- Deletions CACGTTA → CATGTTA Insertions
  - **CTCGTTA** Mutation CACGTTA CACGTTA Mutation CATGTTA CACTGTA

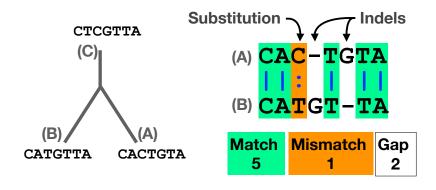
# Alignment view

Alignments are great tools to visualize sequence similarity and evolutionary changes in homologous sequences.

- **Mismatches** represent mutations/substitutions

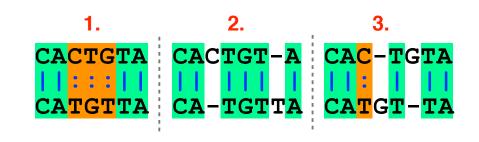
CATGTTA

Gaps represent insertions and deletions (indels)



# Alternative alignments

- Unfortunately, finding the correct alignment is difficult if we do not know the evolutionary history of the two sequences
  - **Q.** Which of these 3 possible alignments is best?



# Alternative alignments

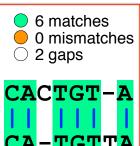
- One way to judge alignments is to compare their number of matches, insertions, deletions and mutations
- 4 matches 3 mismatches O gaps
- 6 matches 0 mismatches O 2 gaps
- 5 matches 1 mismatches 2 gaps

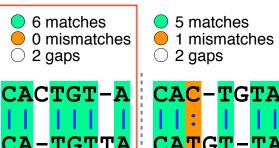




# Optimal alignments

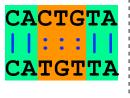
- Biologists often prefer parsimonious alignments, where the number of postulated sequence changes is minimized.
- 4 matches 3 mismatches O gaps





# Scoring alignments

- We can assign a score for each match (+3), mismatch (+1) and indel (-1) to identify the optimal alignment for this scoring scheme
- 0 4 (+3) 0 3 (+1)  $\bigcirc$  0 (-1) = 15
- 6 (+3) 0 (+1)  $\bigcirc$  2 (-1) = 16
- 5 (+3) 1 (+1)  $\bigcirc$  2 (-1) = 14



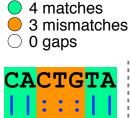


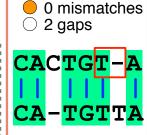


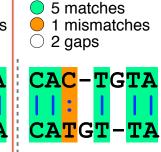
# Optimal alignments

 Biologists often prefer parsimonious alignments, where the number of postulated sequence changes is minimized.

6 matches







# Optimal alignments

- Biologists often prefer parsimonious alignments, where the number of postulated sequence changes is minimized.
- 4 matches3 mismatches0 gaps
- CA<mark>CTG</mark>TA ||:::|| CATGTTA
- 6 matches
  0 mismatches
  2 gaps

  CACTG-TA

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# Optimal alignments

• Biologists often prefer parsimonional alignments, where the numbers of sequence changes is more than one sequences!

4 matches may be more than ot reflect sequences!

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4 matches may be more than ot reflect sequences!

1 matches 1 mismatches 2 gaps

1 mismatches 2 gaps

1 mismatches 2 gaps

1 mismatches 2 gaps

1 cac-TGTA

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- How do we compute the optimal alignment between two sequences?
- Всдот пешпый арргоаст

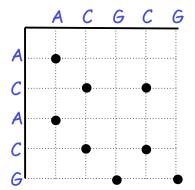
# Dot plots: simple graphical approach

 Place one sequence on the vertical axis of a 2D grid (or matrix) and the other on the horizontal

A C G C G
A
C
A
C

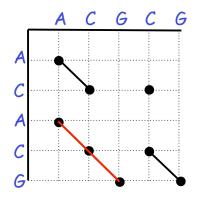
# Dot plots: simple graphical approach

 Now simply put dots where the horizontal and vertical sequence values match



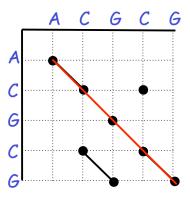
# Dot plots: simple graphical approach

 Diagonal runs of dots indicate matched segments of sequence



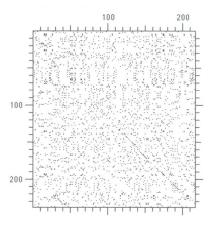
# Dot plots: simple graphical approach

**Q.** What would the dot matrix of a two identical sequences look like?



# Dot plots: simple graphical approach

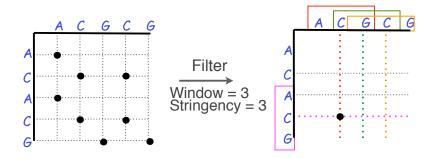
Dot matrices for long sequences can be noisy



# Dot plots: window size and match stringency

Solution: use a window and a threshold

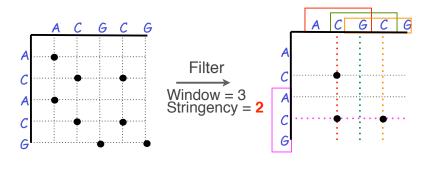
- compare character by character within a window
- require certain fraction of matches within window in order to display it with a dot.
  - You have to choose window size and stringency



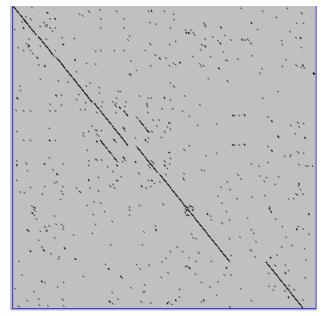
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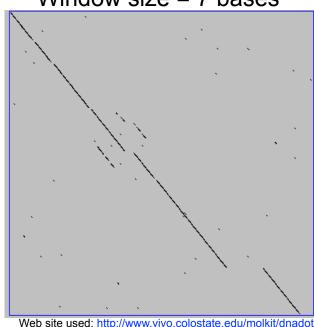
## Window size = 5 bases



A dot plot simply puts a dot where two sequences match. In this example, dots are placed in the plot if 5 bases in a row match perfectly. Requiring a 5 base perfect match is a heuristic – only look at regions that have a certain degree of identity.

Do you expect evolutionarily related sequences to have more word matches (matches in a row over a certain length) than random or unrelated sequences?

#### Window size = 7 bases

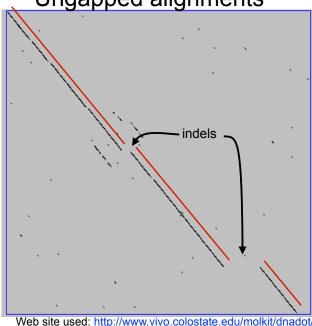


This is a dot plot of the same sequence pair. Now 7 bases in a row must match for a dot to be place. Noise is reduced.

Using windows of a certain length is very similar to using words (kmers) of N characters in the heuristic alignment search tools

Bigger window (kmer) fewer matches to consider

**Ungapped alignments** 



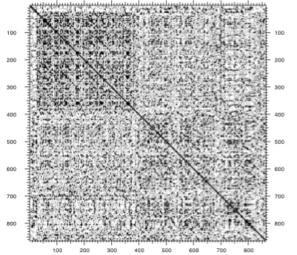
Only **diagonals** can be followed.

Downward or rightward paths represent **insertion** or **deletions** (gaps in one sequence or the other).

#### Uses for dot matrices

- Visually assessing the similarity of two protein or two nucleic acid sequences
- Finding local repeat sequences within a larger sequence by comparing a sequence to itself
  - Repeats appear as a set of diagonal runs stacked vertically and/or horizontally

# Repeats



Human LDL receptor protein sequence (Genbank P01130)

W = 1

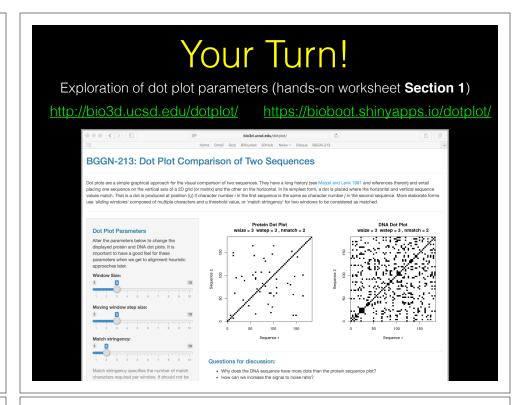
S = 1

(Figure from Mount, "Bioinformatics sequence and genome analysis")

# Repeats Human LDL receptor protein sequence (Genbank P01130) W = 23 S = 7 (Figure from Mount, "Bioinformatics sequence and genome analysis")

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# The Dynamic Programming Algorithm

- The dynamic programming algorithm can be thought of an extension to the dot plot approach
  - One sequence is placed down the side of a grid and another across the top
  - Instead of placing a dot in the grid, we **compute a score** for each position
  - Finding the optimal alignment corresponds to finding the path through the grid with the best possible score



**Needleman, S.B. & Wunsch, C.D.** (1970) "A general method applicable to the search for similarities in the amino acid sequences of two proteins." J. Mol. Biol. 48:443-453.

# Algorithm of Needleman and Wunsch

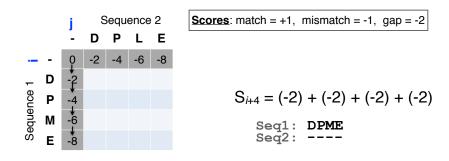
- The Needleman–Wunsch approach to global sequence alignment has three basic steps:
  - (1) setting up a 2D-grid (or alignment matrix),
  - (2) scoring the matrix, and
  - (3) identifying the **optimal path** through the matrix



**Needleman, S.B. & Wunsch, C.D.** (1970) "A general method applicable to the search for similarities in the amino acid sequences of two proteins." J. Mol. Biol. 48:443-453.

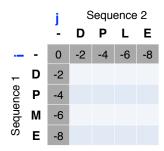
# Scoring the alignment matrix

- Start by filling in the first row and column these are all indels (gaps).
  - Each step you take you will add the **gap penalty** to the score  $(S_{i,j})$  accumulated in the previous cell



# Scoring the alignment matrix

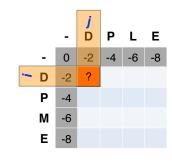
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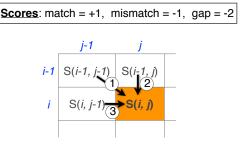


Scores: match = +1, mismatch = -1, gap = -2

# Scoring the alignment matrix

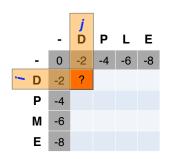
- Then go to the empty corner cell (upper left). It has filled in values in up, left and diagonal directions
  - Now can ask which of the three directions gives the highest score?
  - keep track of this score and direction



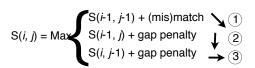


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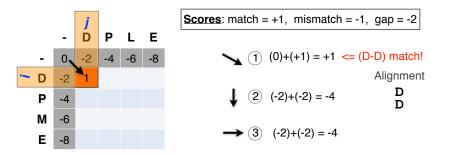


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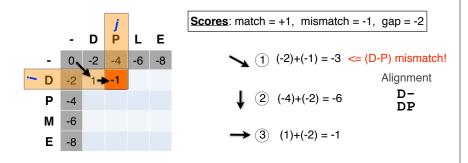
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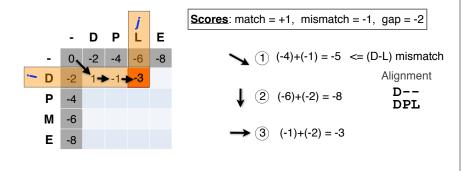
# Scoring the alignment matrix

- At each step, the score in the current cell is determine by the scores in the neighboring cells
  - The maximal score and the direction that gave that score is stored (we will use these later to determine the optimal alignment)



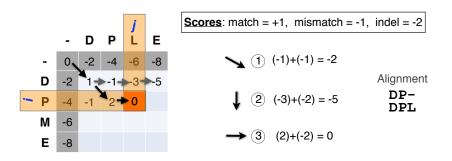
# Scoring the alignment matrix

• We will continue to store the alignment score (S<sub>i,j</sub>) for all possible alignments in the alignment matrix.



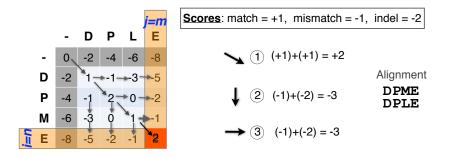
# Scoring the alignment matrix

• For the highlighted cell, the corresponding score (S<sub>i,j</sub>) refers to the score of the optimal alignment of the first *i* characters from sequence1, and the first *j* characters from sequence2.



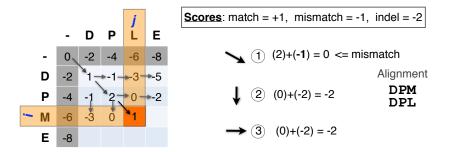
# Scoring the alignment matrix

- The score of the best alignment of the entire sequences corresponds to  $S_{n,m}$ 
  - (where n and m are the length of the sequences)



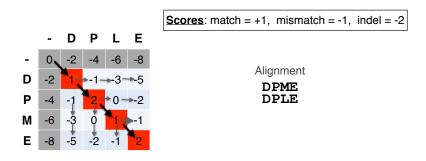
# Scoring the alignment matrix

- At each step, the score in the current cell is determine by the scores in the neighboring cells
  - The maximal score and the direction that gave that score is stored



# Scoring the alignment matrix

- To find the best alignment, we retrace the arrows starting from the bottom right cell
  - N.B. The optimal alignment score and alignment are dependent on the chosen scoring system



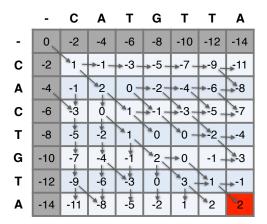
#### Questions:

 What is the optimal score for the alignment of these sequences and how do we find the optimal alignment?

	-	С	Α	Т	G	Т	Т	Α
-	0	-2	-4	-6	-8	-10	-12	-14
С	-2	×1 -	<b>→-1</b> –	<b>→</b> -3_	<b>→</b> -5 –	<b>→</b> -7 –	<b>→</b> -9_	<b>₌</b> 11
A	-4	-1	2	0_	<b>-2</b>	<b>→-4</b>	<b>-</b> -6	-8
С	-6	-3	Ŏ	×1-	<del>)</del> -1-	-3-	-5 _	-7
Т	-8	-5	-2	1	<b>*</b> 0	<b>~</b> 0 -	-2-	<b>→</b> -4
G	-10	-7	-4	-1	2-	<b>→</b> 0	-1 -	-3
Т	-12	-9	-6	-3	Ŏ	3 -	<del>,</del> 1 ←	<b>-</b> -1
A	-14	-11	-8	-5	-2	Ť	2	2

#### Questions:

 What is the optimal score for the alignment of these sequences and how do we find the optimal alignment?



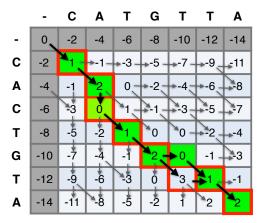
#### Questions:

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	-	С	Α	Т	G	Т	T	Α
-	0	-2	-4	-6	-8	-10	-12	-14
С	-2	1 -	<b>&gt;</b> -1–	<b>→</b> -3_	<b>→</b> -5 –	<b>→</b> -7 –	<b>→</b> -9_	<u></u> -11
A	-4	-1	2	0_	<b>-2</b>	<b>→-4</b>	<b>-</b> -6	-8
С	-6	-3	Ŏ.	×1,-	<del>)</del> -1-	-3-	-5 _	-7
T	-8	-5	-2	1	<b>*</b> 0	×0 -	-2-	<b>→</b> -4
G	-10	-7	-4	-1	2-	<b>→</b> 0 <	-1 -	-3
Т	-12	-9	-6	-3	Ŏ	3 -	<b>3</b> 1 5	<b>⊳</b> -1
A	-14	-11	-8	-5	-2	1	2	2

# More than one alignment possible

 Sometimes more than one alignment can result in the same optimal score

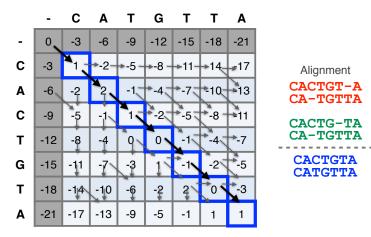


Alignment
CACTGT-A
CA-TGTTA

CACTG-TA CA-TGTTA

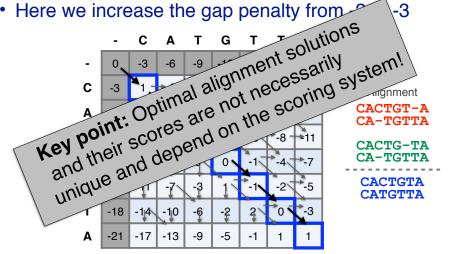
# The alignment and score are dependent on the scoring system

Here we increase the gap penalty from -2 to -3



# The alignment and score are dependent on the scoring system

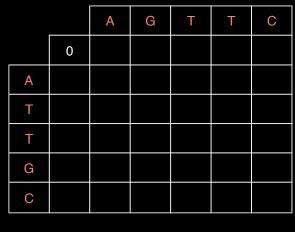
Here we increase the gap penalty from



# Your Turn!

Hands-on worksheet Sections 2 & 3

Match: +2 Mismatch: - I Gap: -2

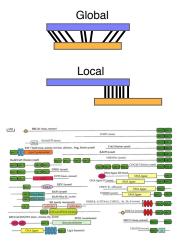


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## Global vs local alignments

- Needleman-Wunsch is a global alignment algorithm
  - Resulting alignment spans the complete sequences end to end
  - This is appropriate for closely related sequences that are similar in length
- For many practical applications we require local alignments
  - Local alignments highlight subregions (e.g. protein domains) in the two sequences that align well



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# Local alignment: Definition

 Smith & Waterman proposed simply that a local alignment of two sequences allow arbitrary-length segments of each sequence to be aligned, with no penalty for the unaligned portions of the sequences.
 Otherwise, the score for a local alignment is calculated the same way as that for a global alignment

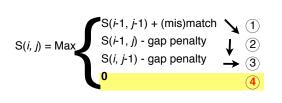
Smith, T.F. & Waterman, M.S. (1981) "Identification of common molecular subsequences." J. Mol. Biol. 147:195-197.

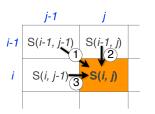
Sequence 1

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# The Smith-Waterman algorithm

- Three main modifications to Needleman-Wunsch:
  - Allow a node to start at 0
  - The score for a particular cell cannot be negative
    - if all other score options produce a negative value, then a zero must be inserted in the cell
  - Record the highest- scoring node, and trace back from there



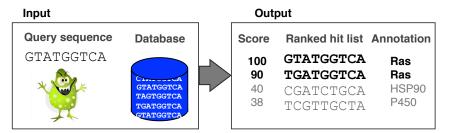


C A G C C U C G C U U A G 0.0 0.0 0.0 0.7 0.3 0.0 1.0 0.0 0.0 0.0 1.0 1.0 0.0 0.7 0.0 0.0 0.0 1.0 0.3 0.0 0.0 0.7 1.0 0.0 0.0 0.7 0.7 1.0 0.0 1.0 0.0 0.0 2.0 1.3 0.3 1.0 0.3 2.0 0.7 0.3 0.3 0.3 0.0 1.0 0.7 0.0 1.0 3.0 1.7 1.3 1.0 1.3 1.7 0.3 0.0 0.0 0.0 0.0 2.0 0.7 0.3 1.7 2.7 1.3 1.0 0.7 1.0 1.3 1.3 0.0 0.0 0.0 0.7 1.7 0.3 1.3 2.7 2.3 1.0 0.7 1.7 2.0 1.0 1.0 0.0 0.0 0.3 0.3 1.3 1.0 2.3 2.3 2.0 0.7 1.7 2.7 1.7 1.0 0.0 0.0 0.0 1.3 0.0 1.0 1.0 2.0 3.3 2.0 1.7 1.3 2.3 2.7 0.0 0.0 1.0 0.0 1.0 0.3 0.7 0.7 2.0 3.0 1.7 1.3 2.3 2.0 0.0 1.0 0.0 0.7 1.0 2.0 0.7 1.7 1.7 3.0 2.7 1.3 1.0 2.0 0.0 0.0 0.7 1.0 0.3 0.7 1.7 0.3 2.7 1.7 2.7 2.3 1.0 2.0 0.0 0.0 0.0 1.7 0.7 0.3 0.3 1.3 1.3 2.3 1.3 2.3 2.0 2.0

Local alignment
GCC-AUG
GCCUCGC

# Local alignments can be used for database searching

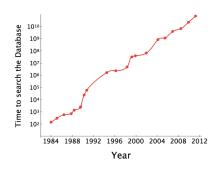
- Goal: Given a guery sequence (Q) and a sequence database (D), find a list of sequences from D that are most similar to Q
  - Input: Q, D and scoring scheme
  - Output: Ranked list of hits



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# The database search problem

- Due to the rapid growth of sequence databases, search algorithms have to be both efficient and sensitive
  - Time to search with SW is proportional to  $m \times n$  (m is length of query. n is length of database), too slow for large databases!

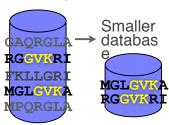


To reduce search time heuristic algorithms. such as BLAST, first remove database sequences without a strong local similarity to the query sequence in a quick initial scan.

# The database search problem

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#### Query RGGVKRIKLMR



To reduce search time heuristic algorithms. such as BLAST, first remove database sequences without a strong local similarity to the query sequence in a quick initial scan.

# ALIGNMENT FOUNDATIONS

- Whv...
  - Why compare biological sequences?
- What...
  - Alignment view of sequence changes during evolution (matches, mismatches and gaps)
- How...
  - Dot matrices
  - Dynamic programing
    - Global alignment
    - Local alignment
  - BLAST heuristic approach

# Rapid, heuristic versions of Smith-Waterman: BLAST

- BLAST (Basic Local Alignment Search Tool) is a simplified form of Smith-Waterman (SW) alignment that is popular because it is fast and easily accessible
  - BLAST is a heuristic approximation to SW It examines only part of the search space
  - BLAST saves time by restricting the search by scanning database sequences for likely matches before performing more rigorous alignments
  - Sacrifices some sensitivity in exchange for speed
  - In contrast to SW, BLAST is not guaranteed to find optimal alignments

Rapid, heuristic versions of Smith-Waterman: BLAST

The central idea of the BLAST algorithm is to confine attention an initial word nair match.

The sequence nairs that contain an initial word nair match.

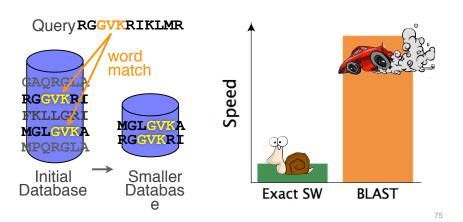
"The central idea of the BLAST algorithm is to confine atter to sequence pairs that contain an initial word pair match"

to sequence pairs that (100n)

ast to SW, BLAST is not guaranteed to find optimal andnments

ıat

 BLAST uses this pre-screening heuristic approximation resulting in an an approach that is about 50 times faster than the Smith-Waterman



#### How BLAST works

Four basic phases

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- Phase 1: compile a list of guery word pairs (w=3)

RGGVKRI Query sequence RGG **GGV GVK** generate list VKR of w=3words for KRI query

#### Blast

 Phase 2: expand word pairs to include those similar to query (defined as those above a similarity threshold to original word, i.e. match scores in substitution matrix)

```
RGGVKRI Query sequence
RGG RAG RIG RLG ...
GGV GAV GTV GCV ...

extend list of GVK GAK GIK GGK ...
words similar VKR VRR VHR VER ...
to query KRI KKI KHI KDI ...
```

Blast

 Phase 3: a database is scanned to find sequence entries that match the compiled word list

```
RGGVKRI Query sequence

RGG RAG RIG RLG . . .

search for GGV GAV GTV GCV . . .

perfect GVK GLK GIK GGK . . .

matches in the database kRI KKI KHI KDI . . .

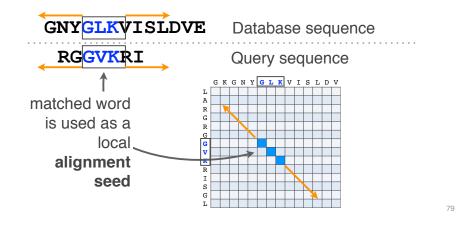
sequence
```

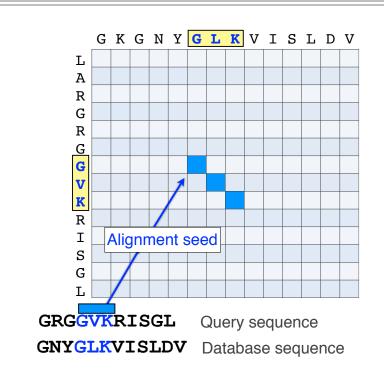
78

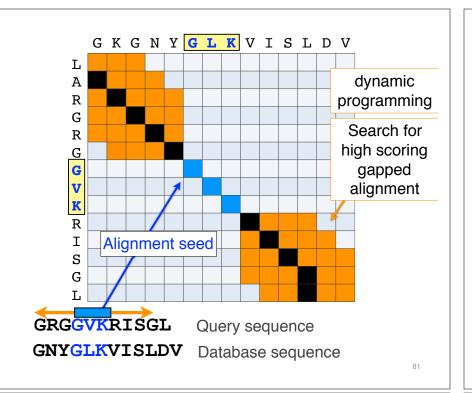
#### **Blast**

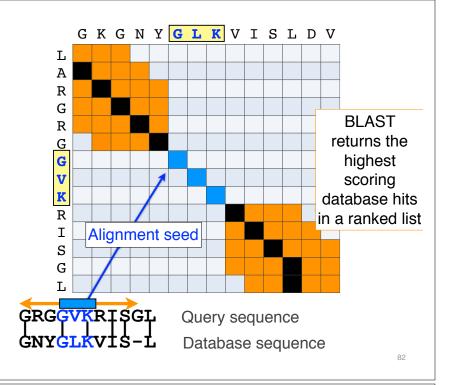
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 Phase 4: the initial database hits are extended in both directions using dynamic programing









# **BLAST** output

 BLAST returns the highest scoring database hits in a ranked list along with details about the target sequence and alignment statistics

Description	Max score	Query cover	E value	Max ident	Accession	
kinesin-1 heavy chain [Homo sapiens]	677	100%	0	100%	NP_004512.1	
Kif5b protein [Mus musculus]	676	100%	0	98%	AAA20133.1	
Kinesin-14 heavy chain [Danio rerio]	595	88%	0	78%	XP_00320703	
hypothetical protein EGK_18589	48.2	40%	0.03	32%	ELK35081.1	
mKIAA4102 protein [Mus musculus]	42.7	38%	3.02	24%	EHH28205.1	

# Statistical significance of results

 An important feature of BLAST is the computation of statistical significance for each hit.
 This is described by the **E value** (expect value)

Description	Max score	cover	E value	Max ident	Accession
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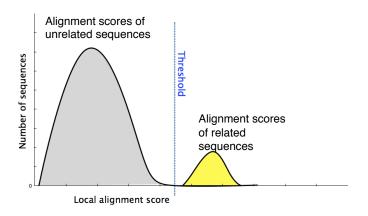
#### **BLAST scores and E-values**

- The E value is the expected number of hits that are as good or better than the observed local alignment score (with this score or better) if the query and database are random with respect to each other
  - -i.e. the number of alignments expected to occur by chance with equivalent or better scores
- Typically, only hits with E value below a significance threshold are reported
  - This is equivalent to selecting alignments with score above a certain score threshold

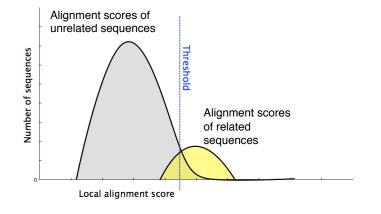
85

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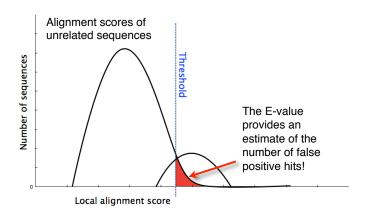
 Ideally, a threshold separates all query related sequences (yellow) from all unrelated sequences (gray)

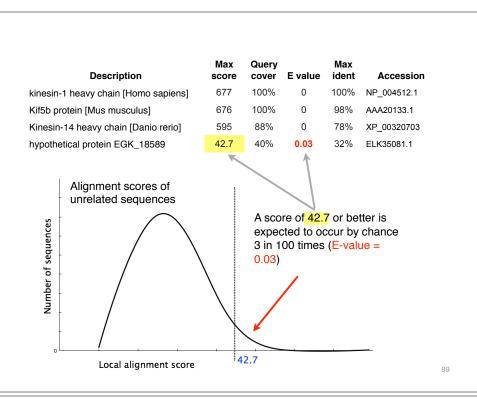


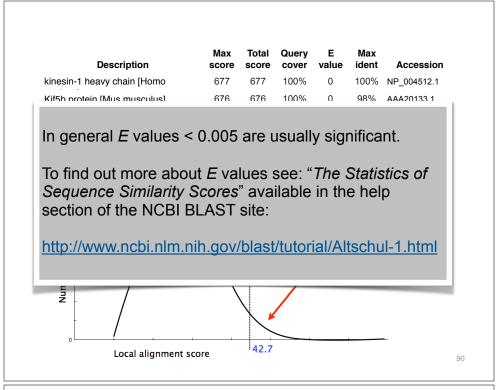
- Unfortunately, often both score distributions overlap
  - The E value describes the expected number of hits with a score above the threshold if the query and database are unrelated



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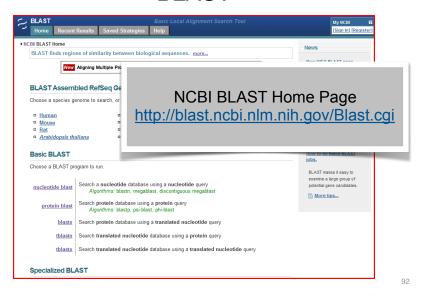


# Your Turn!

Hands-on worksheet Sections 4 & 5

- ▶ Please do answer the last lab review question (Q19).
- We encourage discussion and exploration!

# Practical database searching with BLAST



# Practical database searching with BLAST

- There are four basic components to a traditional BLAST search
  - (1) Choose the sequence (query)
  - (2) Select the BLAST program
  - (3) Choose the database to search
  - (4) Choose optional parameters
- Then click "BLAST"

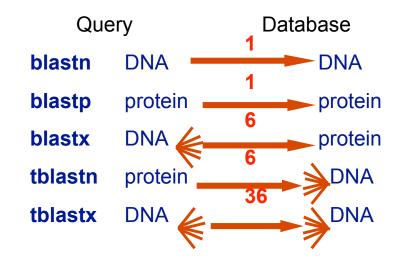
Step 1: Choose your sequence

 Sequence can be input in FASTA format or as accession number

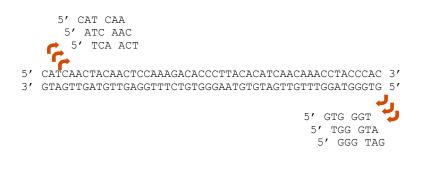


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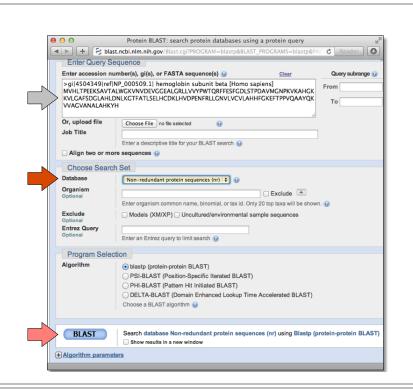
# **Step 2**: Choose the BLAST program



# DNA potentially encodes six proteins

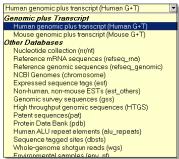


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# **Step 3**: Choose the database

nr = non-redundant (most general database) dbest = database of expressed sequence tags dbsts = database of sequence tag sites gss = genomic survey sequences



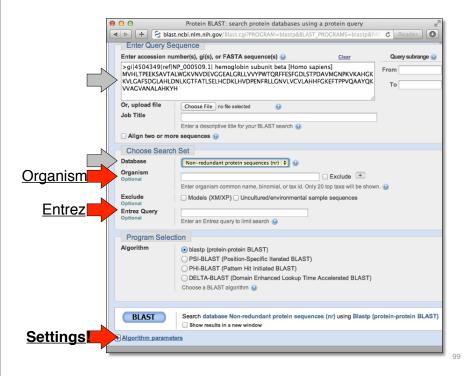
Non-redundant protein sequences (nr)
Non-redundant protein sequences (nr)
Reference proteins (refseq\_protein)
Swissprot protein sequences(swissprot)
Patented protein sequences(pat)
Protein Data Bank proteins(pdb)
Environmental semples(env\_nr)

protein databases

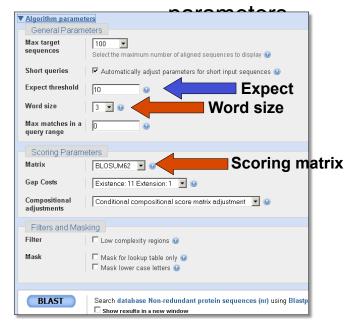
nucleotide databases

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# Step 4a: Select optional search

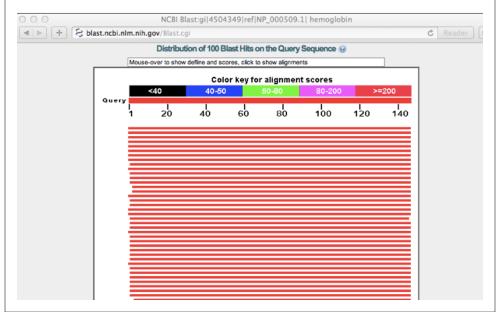


# Step 4: Optional parameters

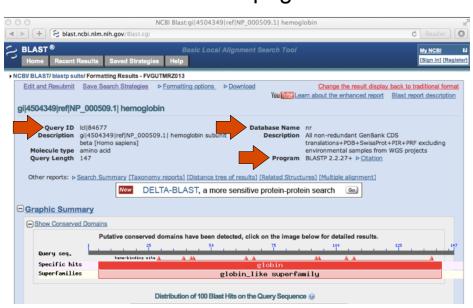
- You can...
  - choose the organism to search
  - change the substitution matrix
  - change the expect (E) value
  - change the word size
  - change the output format

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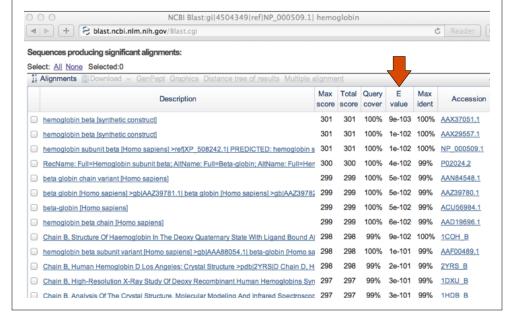
# Further down the results page...



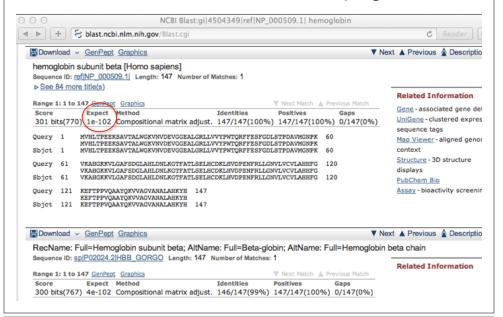
# Results page



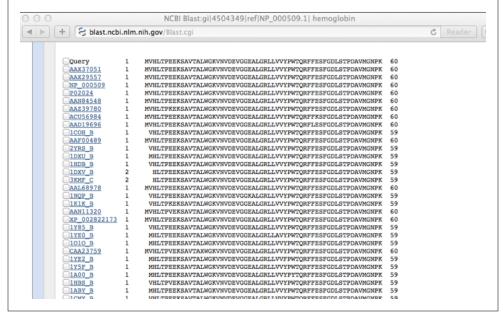
# Further down the results page...



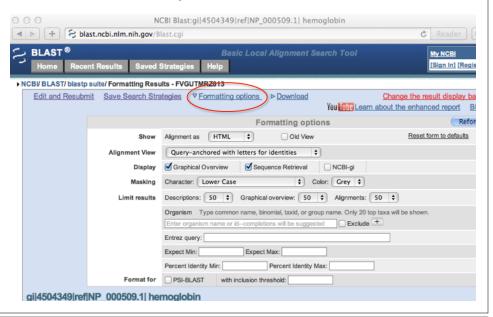
# Further down the results page...



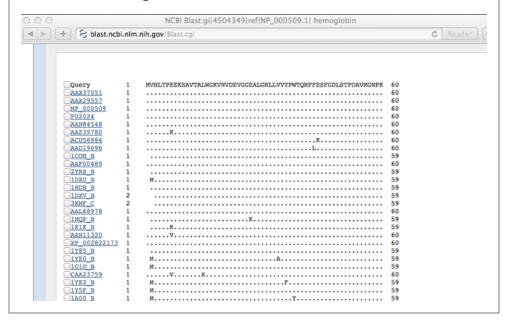
# E.g. Query anchored alignments



# Different output formats are available



# ... and alignments with dots for identities



## Common problems

- Selecting the wrong version of BLAST
- Selecting the wrong database
- Too many hits returned
- Too few hits returned
- Unclear about the significance of a particular result - are these sequences homologous?

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# How to handle too many results

- Focus on the question you are trying to answer
  - select "refseq" database to eliminate redundant matches from "nr"
  - Limit hits by organism
  - Use just a portion of the query sequence, when appropriate
  - Adjust the expect value; lowering E will reduce the number of matches returned

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#### How to handle too few results

- Many genes and proteins have no significant database matches
  - remove Entrez limits
  - raise E-value threshold
  - search different databases
  - try scoring matrices with lower BLOSUM values (or higher PAM values)
  - use a search algorithm that is more sensitive than BLAST (e.g. PSI-BLAST or HMMer)

# Summary of key points

- Sequence alignment is a fundamental operation underlying much of bioinformatics.
- Even when optimal solutions can be obtained they are not necessarily unique or reflective of the biologically correct alignment.
- Dynamic programming is a classic approach for solving the pairwise alignment problem.
- Global and local alignment, and their major application areas.
- Heuristic approaches are necessary for large database searches and many genomic applications.

# FOR NEXT CLASS...

#### Check out the online:

**Reading**: Sean Eddy's "What is dynamic programming?"

Homework: (1) Quiz, (2) Alignment Exercise.

Update!

# **Homework Grading**

Both (1) quiz questions and (2) alignment exercise carry equal weights (i.e. 50% each).

(Homework 2) Assessment Criteria	Points	
Setup labeled alignment matrix	1	
Include initial column and row for GAPs	1	
All alignment matrix elements scored (i.e. filled in)	1	
Evidence for correct use of scoring scheme	1	
Direction arrows drawn between all cells	1	
Evidence of multiple arrows to a given cell if appropriate	1	D
Correct optimal score position in matrix used	1	С
Correct optimal score obtained for given scoring scheme	1	В
Traceback path(s) clearly highlighted	1	Α
Correct alignment(s) yielding optimal score listed	1	A+