

#### **Recap From Last Time:**

- Bioinformatics is computer aided biology.
  - Deals with the collection, archiving, organization, and interpretation of a wide range of biological data.
- The NCBI and EBI are major online bioinformatics service providers.
- Introduced via hands-on session the BLAST, Entrez, GENE, OMIM, UniProt, Muscle and PDB bioinformatics tools and databases.
  - Muddy point assessment (see <u>results</u>)
- There are a large number of bioinformatics databases (see handout!).
- Also covered: Course structure; Supporting course website, Ethics code, and Introductions...

## Today's Menu

Classifying Databases	Primary, secondary and composite Bioinformatics databases				
Using Databases	Vignette demonstrating how major Bioinformatics databases intersect				
Major Biomolecular Formats	How nucleotide and protein sequence and structure data are represented				
Alignment Foundations	Introducing the <i>why</i> and <i>how</i> of comparing sequences				
Alignment Algorithms	Hands-on exploration of alignment algorithms and applications				

#### Primary, secondary & composite databases

Bioinformatics databases can be usefully classified into *primary*, *secondary* and *composite* according to their data source.

- Primary databases (or <u>archival databases</u>) consist of data derived experimentally.
  - → GenBank: NCBI's primary nucleotide sequence database.
  - ▶ PDB: Protein X-ray crystal and NMR structures.
- Secondary databases (or <u>derived databases</u>) contain information derived from a primary database.
  - RefSeq: non redundant set of curated reference sequences primarily from GenBank
  - PFAM: protein sequence families primarily from UniProt and PDB
- Composite databases (or metadatabases) join a variety of different primary and secondary database sources.
  - OMIM: catalog of human genes, genetic disorders and related literature
  - GENE: molecular data and literature related to genes with extensive links to other databases.

#### DATABASE VIGNETTE

You have just come out a seminar about gastric cancer and one of your co-workers asks:

"What do you know about that 'Kras' gene the speaker kept taking about?"

You have some recollection about hearing of 'Ras' before. How would you find out more?

- Google?
- · Library?
- Bioinformatics databases at NCBI and EBI!

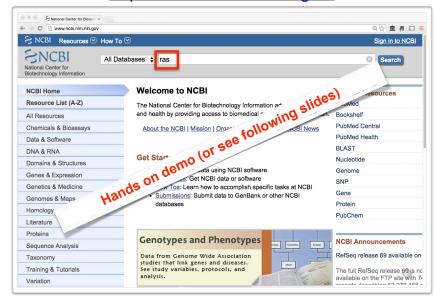
http://www.ncbi.nlm.nih.gov/

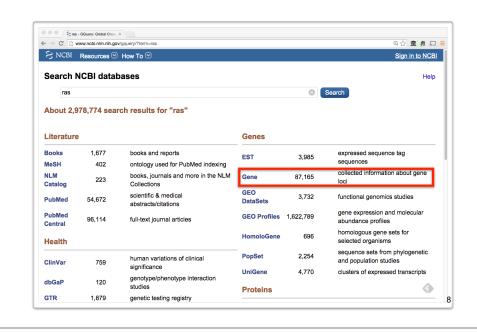
#### **Example Vignette Questions:**

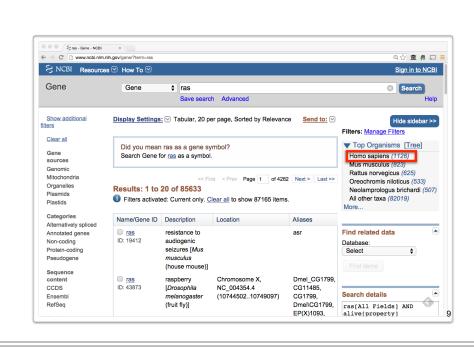
- What chromosome location and what genes are in the vicinity of a given query gene? NCBI GENE
- What can you find out about molecular functions, biological processes, and prominent cellular locations?
- What amino acid positions in the protein are responsible for ligand binding? EBI UniProt
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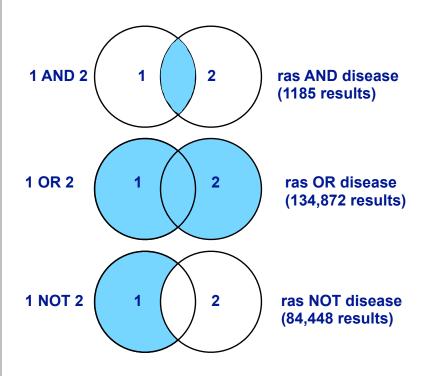
  RCSB PDB

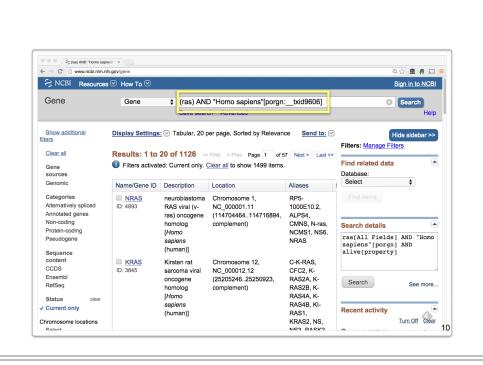
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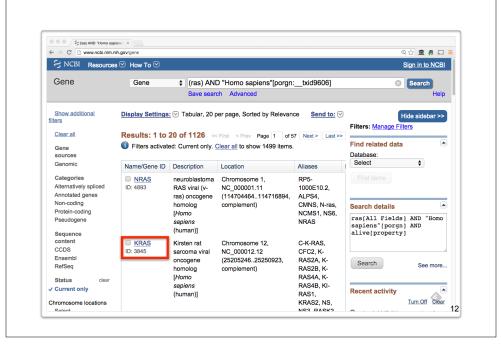


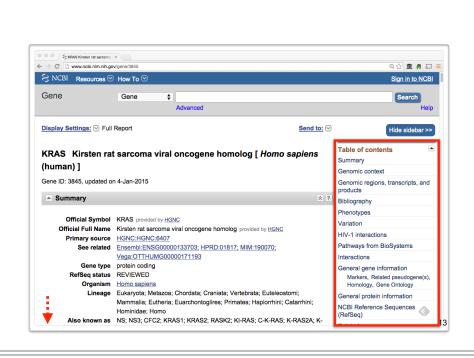


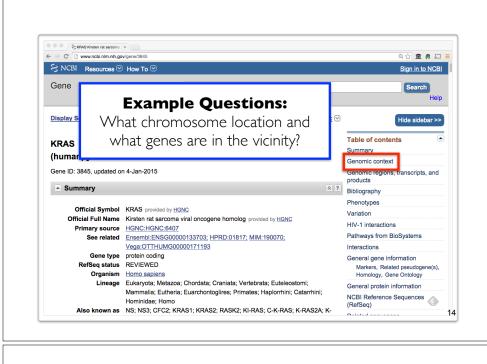


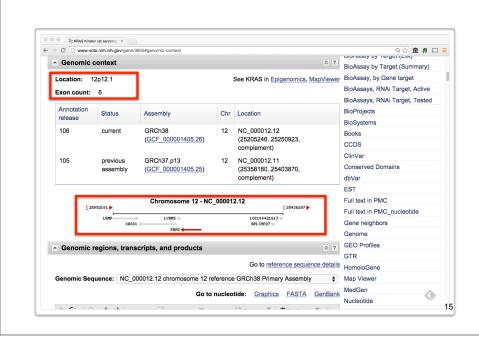


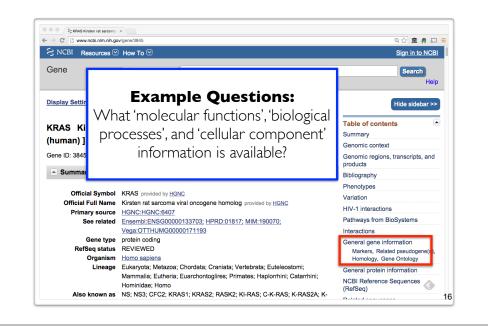


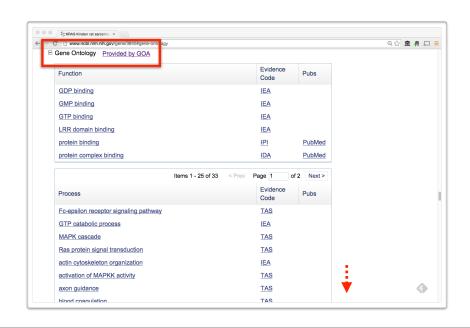






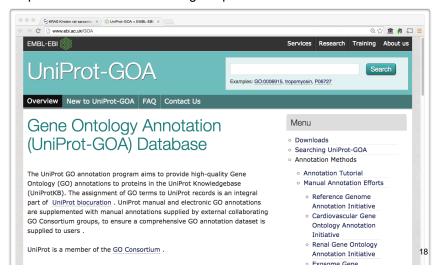






#### **GO: Gene Ontology**

GO provides a controlled vocabulary of terms for describing gene product characteristics and gene product annotation data



#### Why do we need Ontologies?

- Annotation is essential for capturing the understanding and knowledge associated with a sequence or other molecular entity
- Annotation is traditionally recorded as "free text", which is easy to read by humans, but has a number of disadvantages, including:
  - Difficult for computers to parse
  - Quality varies from database to database
  - Terminology used varies from annotator to annotator
- Ontologies are annotations using standard vocabularies that try to address these issues
- GO is integrated with UniProt and many other databases including a number at NCBI

#### **GO Ontologies**

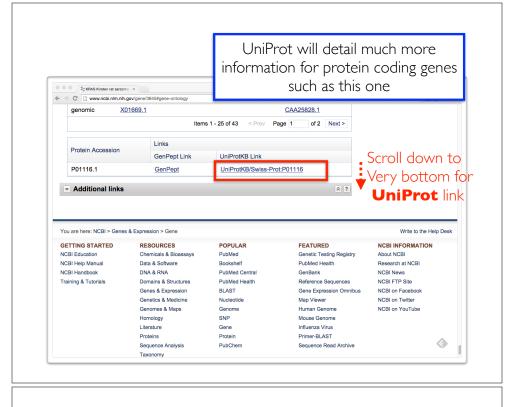
- There are three ontologies in GO:
  - Biological <u>Process</u>
     A commonly recognized series of events e.g. cell division, mitosis,
  - Molecular <u>Function</u>
     An elemental activity, task or job
     e.g. kinase activity, insulin binding
  - Cellular <u>Component</u>
     Where a gene product is located
     e.g. mitochondrion, mitochondrial
     membrane

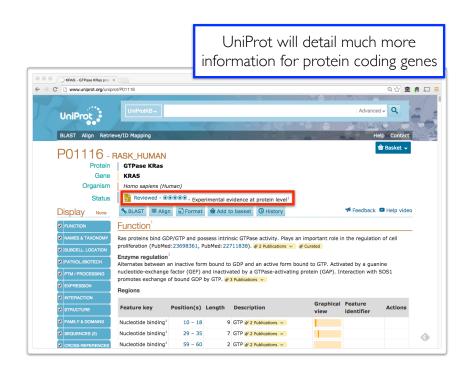


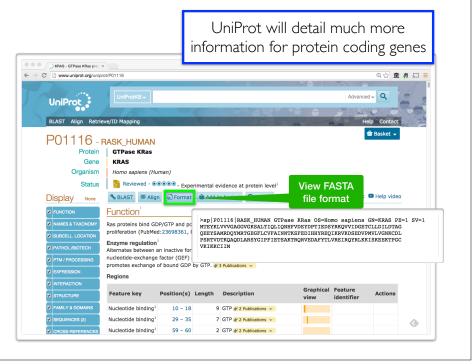


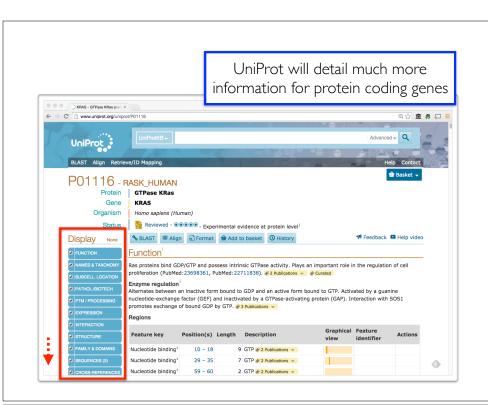


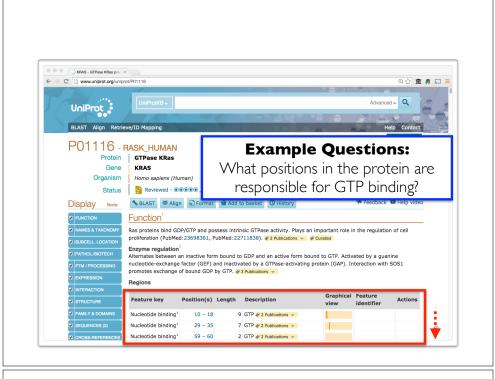


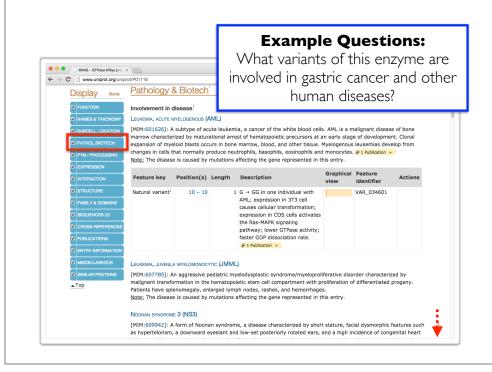


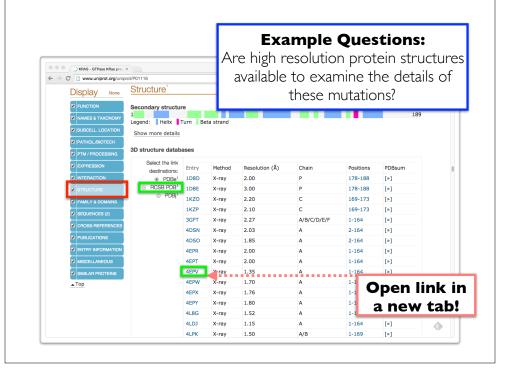


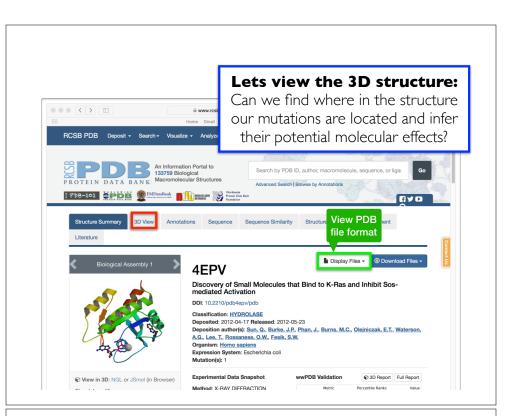


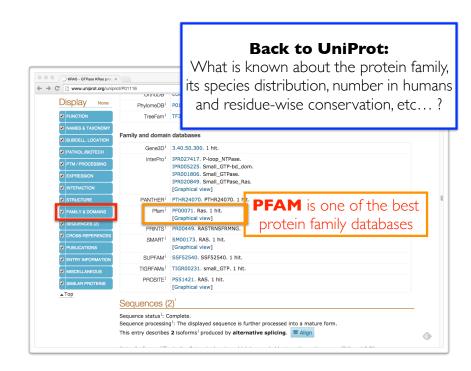


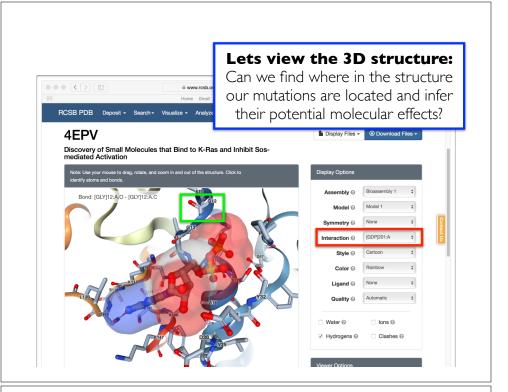


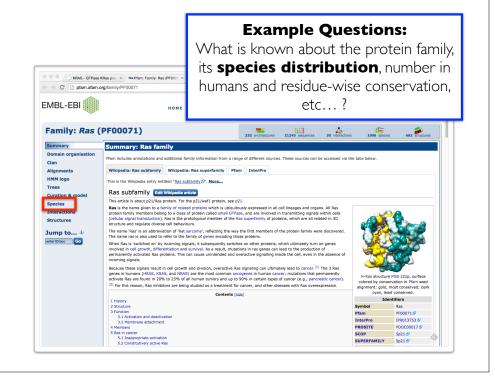


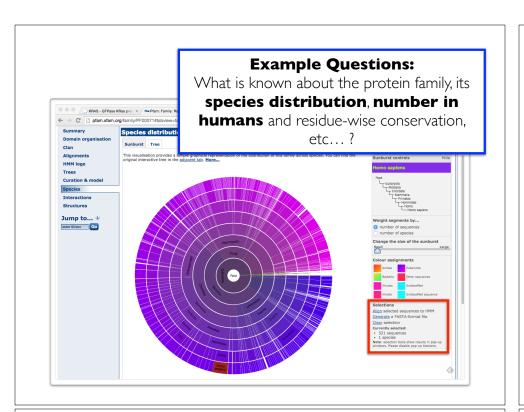


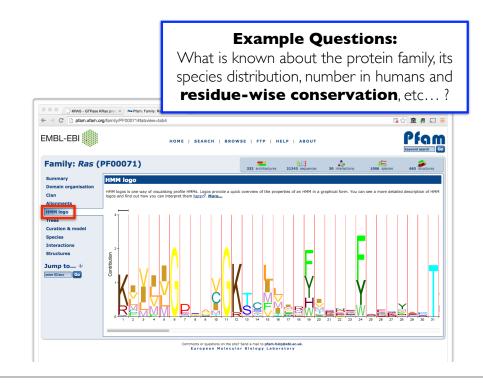


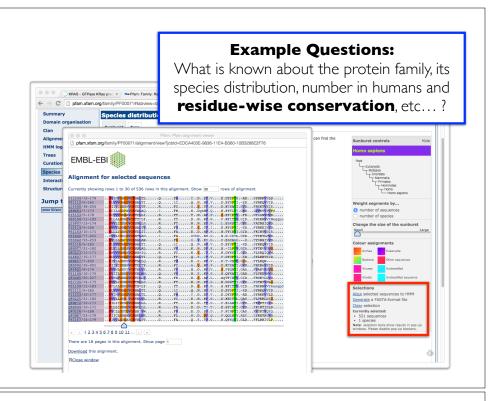


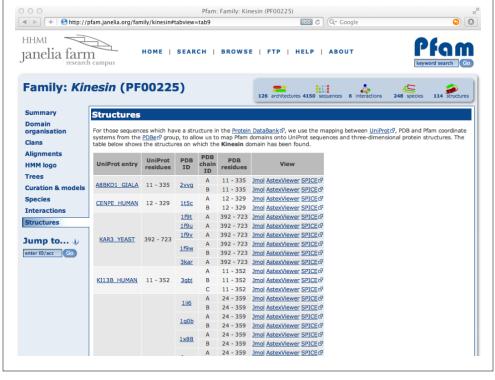












#### Recap: Major NCBI and EBI databases

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  RCSB PDB

#### 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

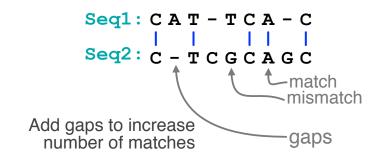
Seq2: CTCGCAGC

[Screencast Material]

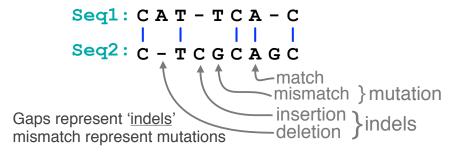
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Two types of character correspondence

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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

- Protein structure prediction

   Assembly of sequence alignment is arguably the such as a bact most fundamental operation of bioinformatics! mg transcription factor binding sites via ChIP-Seq (chromatin immuno-precipitation sequencing)
  - Pretty much all next-gen sequencing data analysis

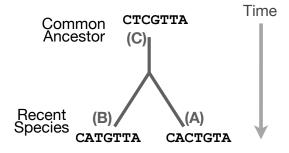
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#### Sequence changes during evolution

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

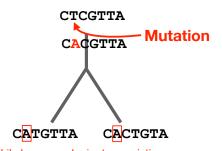
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- Deletions
- Insertions



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- Mutations/Substitutions CTCGTTA → CACGTTA
- Deletions
- Insertions

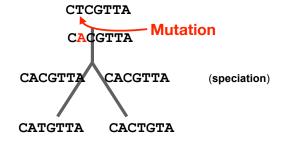


Likely occurred prior to speciation

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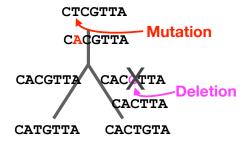
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- Insertions

 $\mathtt{CACGTTA} \longrightarrow \mathtt{CACTTA}$ 



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Mutations/Substitutions
 Deletions
 CTCGTTA → CACGTTA
 CACGTTA → CACTTA

CTCGTTA

- Insertions CACTTA → CACTGTA

CACGTTA CACCTTA Deletion
CACGTTA CACTGTA Insertion

#### Mutations, deletions and insertions

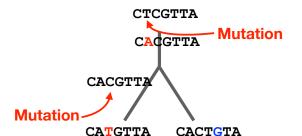
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CTCGTTA → CACGTTA

DeletionsInsertions

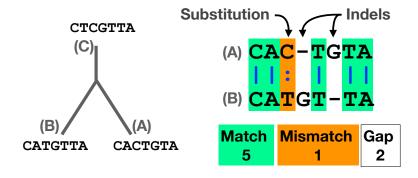
CACGTTA → CATGTTA



#### Alignment view

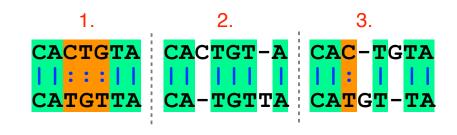
Alignments are great tools to visualize sequence similarity and evolutionary changes in homologous sequences.

- Mismatches represent mutations/substitutions
- Gaps represent insertions and deletions (indels)



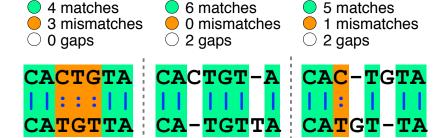
#### Alternative alignments

- 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?



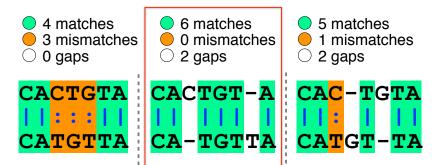
#### Alternative alignments

 One way to judge alignments is to compare their number of matches, insertions, deletions and mutations



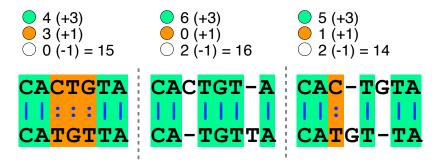
## Optimal alignments

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



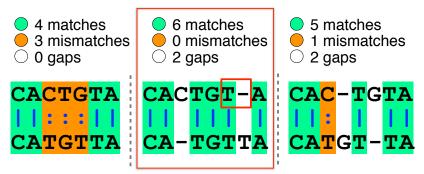
#### Scoring alignments

 We can assign a score for each match (+3), mismatch (+1) and indel (-1) to identify the optimal alignment for this scoring scheme



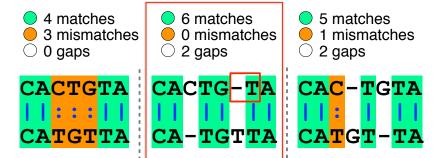
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#### Optimal alignments

alignments, where the number one optimal sequence changes is more than one true sequence changes is more than one true than one optimal sequence changes is more than one optimal sequence than one optimal sequence than one optimal sequence than one optimal sequences!

4 matches may be more than one optimal the true sequences!

3 min There may be may not reflect the true sequences!

1 mismatches
2 gans alignment and these may not reflect the true

any ment and mese may not renect intervences!

evolutionary history of our sequences!

CAC-TGTA

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  - How do we compute the optimal alignment between two sequences?
  - ▶ BLAST HEURSHE APPROACH

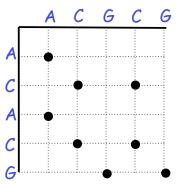
#### Dot plots: simple graphical approach

 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

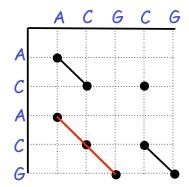
#### Dot plots: simple graphical approach

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



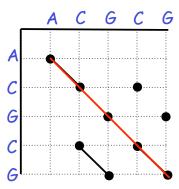
#### Dot plots: simple graphical approach

 Diagonal runs of dots indicate matched segments of sequence



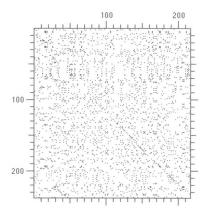
#### Dot plots: simple graphical approach

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



#### Dot plots: simple graphical approach

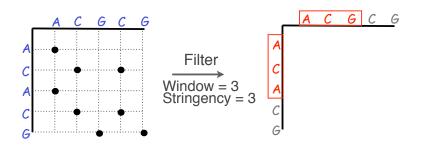
Dot matrices for long sequences can be noisy



# Dot plots: window size and match stringency

Solution: use a window and a threshold

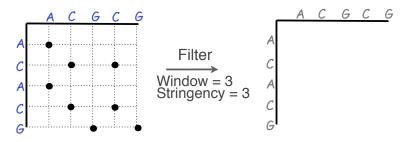
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- require certain fraction of matches within window in order to display it with a dot.
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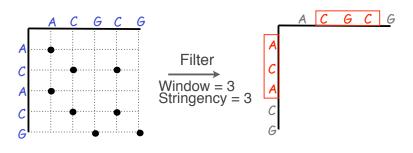
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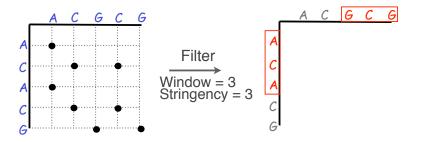
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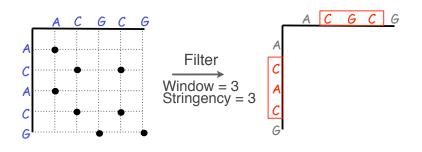
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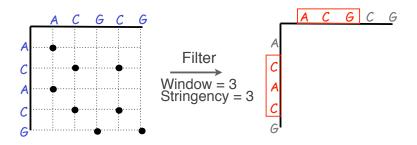
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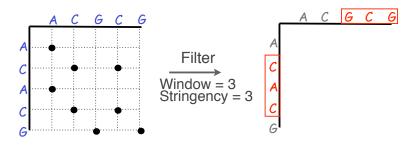
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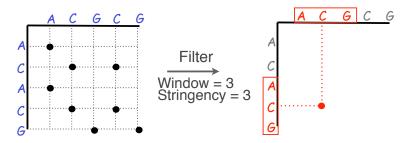
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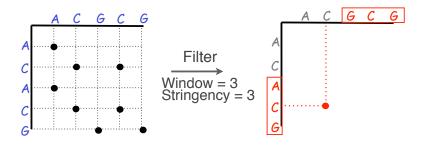
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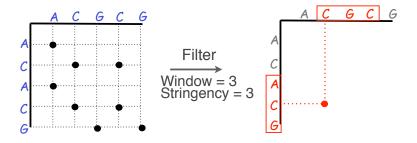
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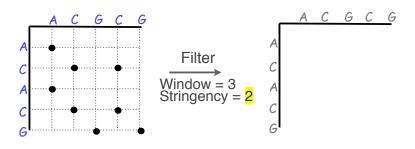
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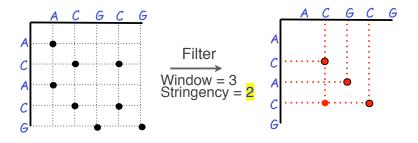
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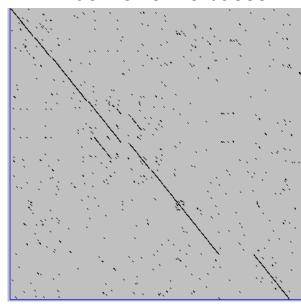


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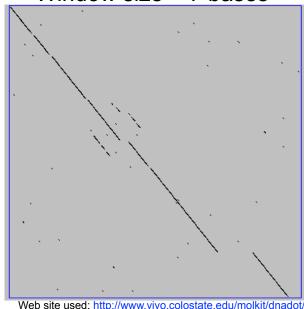
#### Window size = 5 bases



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?

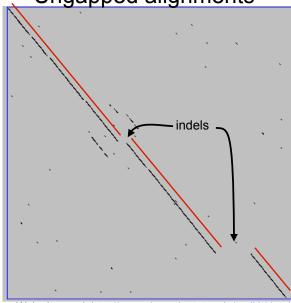
#### Window size = 7 bases



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** 



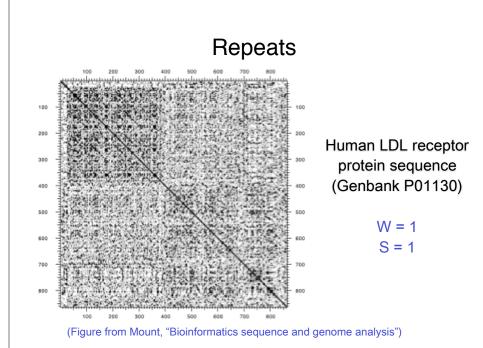
Web site used: http://www.vivo.colostate.edu/molkit/dnado

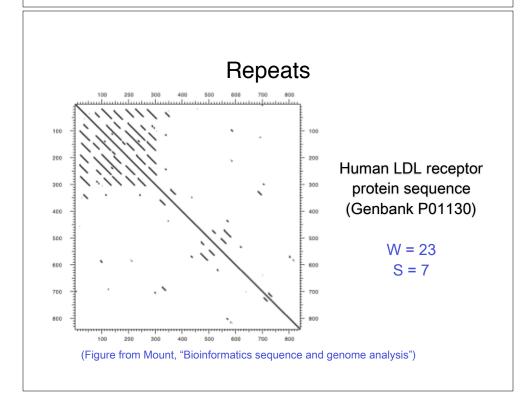
Only diagonals can be followed.

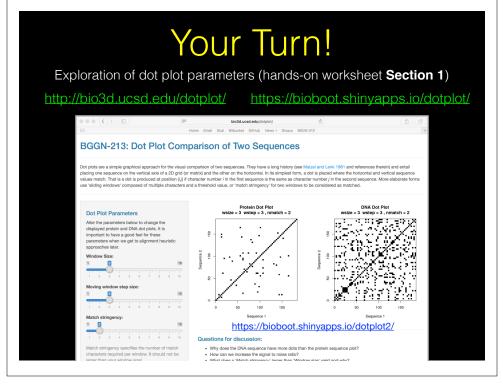
Downward or rightward paths represent insertion or deletions (gaps in one sequence or the other).

#### 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







#### 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

### The Dynamic Programming Algorithm

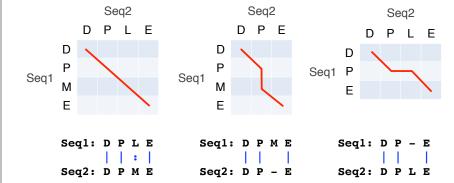
- The dynamic programming algorithm can be thought of an extension to the dot plot approach
  - One sequence is placed down the side of a grid and another across the top
  - Instead of placing a dot in the grid, we compute a score for each position
  - Finding the optimal alignment corresponds to finding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

#### 90

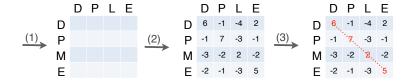
## Different paths represent different alignments



Matches are represented by <u>diagonal paths</u> & indels with horizontal or vertical path segments

#### Algorithm of Needleman and Wunsch

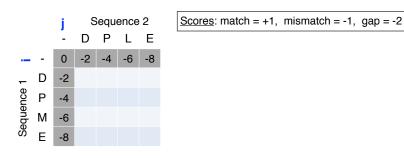
- The Needleman–Wunsch approach to global sequence alignment has three basic steps:
  - (1) setting up a 2D-grid (or alignment matrix),
  - (2) scoring the matrix, and
  - (3) identifying the optimal path through the matrix



**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.

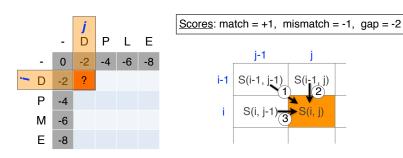
#### Scoring the alignment matrix

- 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



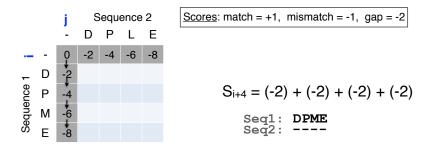
#### Scoring the alignment matrix

- 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



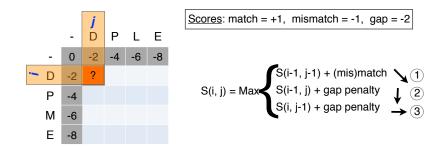
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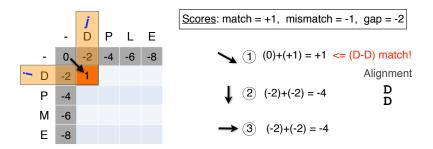
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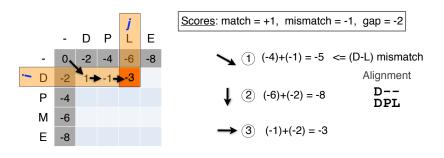
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  - keep track of direction and score



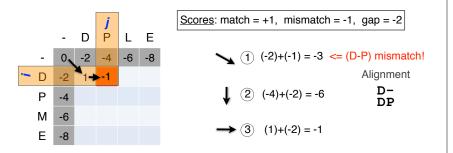
#### Scoring the alignment matrix

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



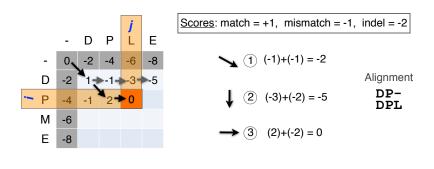
#### Scoring the alignment matrix

- 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)



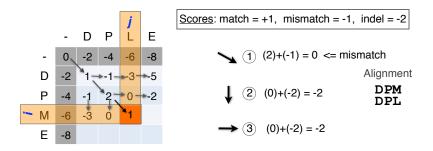
#### Scoring the alignment matrix

• 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.



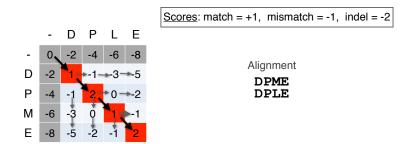
#### Scoring the alignment matrix

- 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



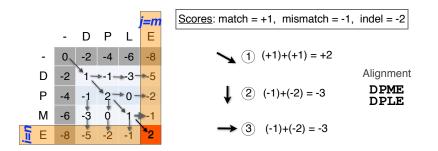
#### Scoring the alignment matrix

- 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



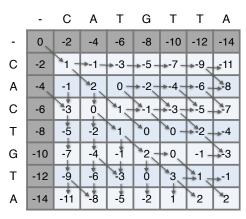
#### Scoring the alignment matrix

- 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)



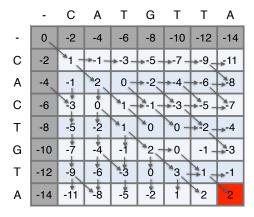
#### Questions:

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



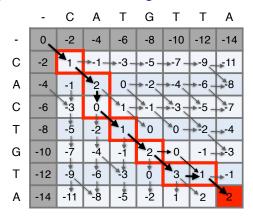
#### **Questions:**

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



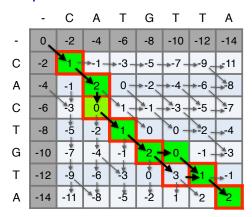
#### Questions:

 To find the best alignment we retrace the arrows starting from the bottom right cell



#### More than one alignment possible

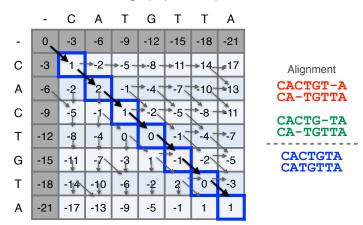
 Sometimes more than one alignment can result in the same optimal score



Alignment
CACTGT-A
CA-TGTTA
CACTG-TA
CA-TGTTA

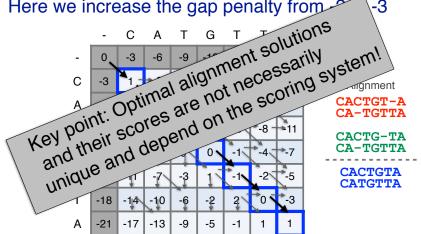
# The alignment and score are dependent on the scoring system

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



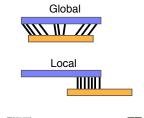
NW [	DYN	AMI	C PF	ROC	SRAI	MMII	NG	
Match: +2			Α	G	Т	Т	С	
Mismatch: - I		0						
Gap: -2	Α							
	Т							
	Т							
	G							
	С							

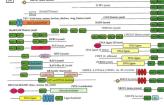
#### ALIGNMENT FOUNDATIONS

- Why...
  - · Why compare biological sequences?
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- How...
  - Dot matrices
  - Dynamic programing
    - Global alignment
    - Local alignment
  - BLAST heuristic approach

#### 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





#### 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.

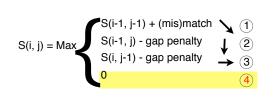
113

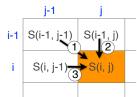
#### Sequence 1 G C C U C G C U U A G 0.0 0.0 1.0 0.7 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1.0 0.7 0.0 0.0 0.0 0.7 0.3 0.0 1.0 0.0 0.0 0.0 1.0 1.0 0.0 0.7 0.0 0.0 0.0 1.0 0.3 0.0 0.0 0.7 1.0 0.0 0.0 0.7 0.7 1.0 0.0 1.0 0.0 0.0 2.0 1.3 0.3 1.0 0.3 2.0 0.7 0.3 0.3 0.3 0.0 1.0 0.7 0.0 1.0 3,0 1.7 1.3 1.0 1.3 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-UCG
GCCAUUG

#### 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

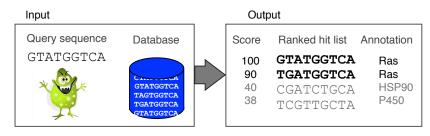




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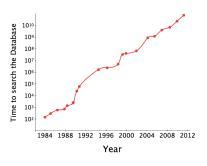
# Local alignments can be used for database searching

- Goal: Given a query 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



#### The database search problem

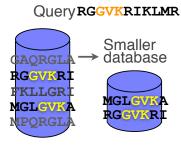
- Due to the rapid growth of sequence databases, search algorithms have to be both efficient and sensitive
  - Time to search with SW is proportional to m x n (m is length of query, n is length of database), too slow for large databases!



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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#### ALIGNMENT FOUNDATIONS

- Why...
  - · Why compare biological sequences?
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- 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

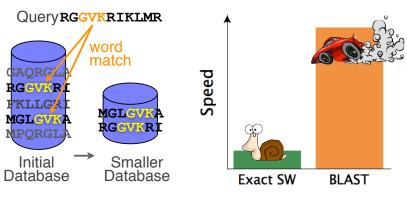
#### Rapid, heuristic versions of Smith-Waterman: BLAST

- "The central idea of the BLAST algorithm is to confine attention

  "The central idea of that contain an initial word pair match"

  to sequence pairs that "The central idea of the BLAST algorithm is to confine atte to sequence pairs that contain an initial word pair match" matches before performing
  - ntrast to SW, BLAST is not guaranteed to find optimal alignments

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



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#### How BI AST works

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

RGGVKRI Query sequence RGG GGV **GVK** generate list

of w=3 words VKR for query KRI

#### **Blast**

- Phase 2: expand word pairs to include those similar to guery (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 GVK GAK GIK GGK extend list of VKR VRR VHR VER words similar to query KRI KKI KHI KDI

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#### 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 ...

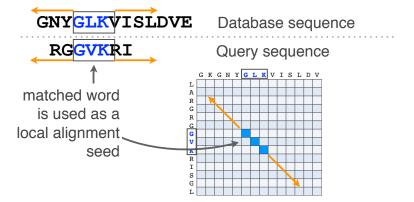
perfect GVK GLK GIK GGK ...

matches in the VKR VRR VHR VER ...

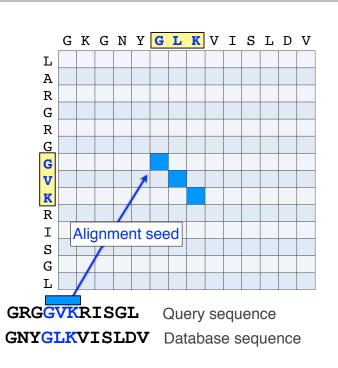
database KRI KKI KHI KDI ...
```

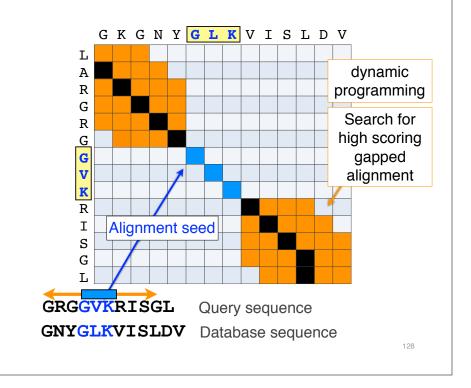
#### Blast

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

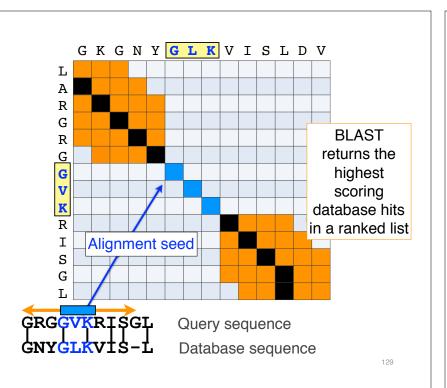


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#### **BLAST** output

 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

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#### Statistical significance of results

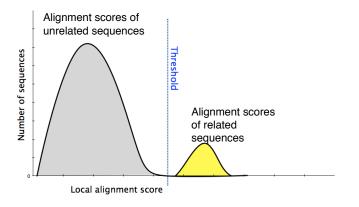
 An important feature of BLAST is the computation of statistical significance for each hit.
 This is described by the E value (expect value)

Description	Max score	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
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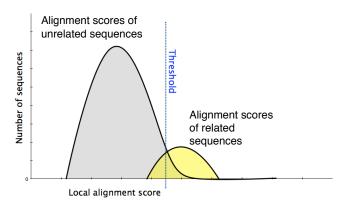
#### **BLAST** scores and E-values

- 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

 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

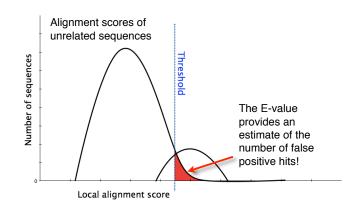


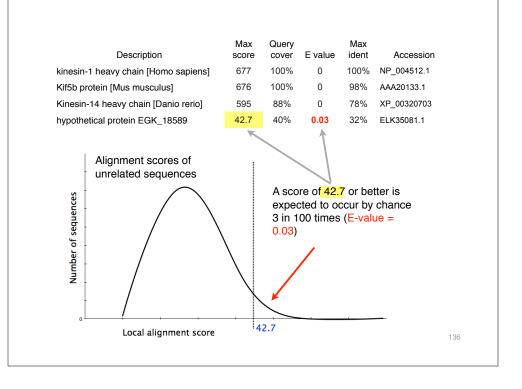
· Unfortunately, often both score distributions overlap

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 The E value describes the expected number of hits with a score above the threshold if the query and database are unrelated



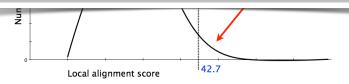


Description	Max score	Total score	Query cover	E value	Max ident	Accession
kinesin-1 heavy chain [Homo	677	677	100%	0	100%	NP_004512.1
Kif5h protein [Mus musculus]	676	676	100%	0	98%	AAA20133.1

In general *E* values < 0.005 are usually significant.

To find out more about *E* values see: "The Statistics of Sequence Similarity Scores" available in the help section of the NCBI BLAST site:

http://www.ncbi.nlm.nih.gov/blast/tutorial/Altschul-1.html



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## Your Turn!

#### Hands-on worksheet **Sections 6**

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

## Summary

- 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.
- Reviewed global and local alignment approaches and their major application areas.
- Heuristic approaches are often 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.

#### **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+

# REFERENCE SLIDES... Additional reference slides for the motivated student

## Practical database searching with BLAST

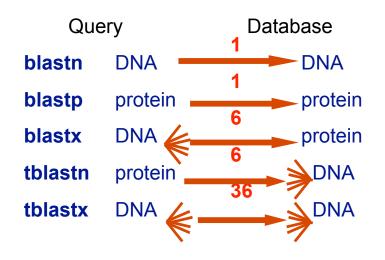
- There are four basic components to a traditional BLAST search
  - (1) Choose the sequence (query)
  - (2) Select the BLAST program
  - (3) Choose the database to search
  - (4) Choose optional parameters
- Then click "BLAST"

#### Step 1: Choose your sequence

 Sequence can be input in FASTA format or as accession number



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



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#### DNA potentially encodes six proteins

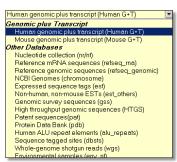
5' CAT CAA
5' ATC AAC
5' TCA ACT
5' CATCAACTACAACTCCAAAGACACCCTTACACATCAACAAACCTACCCAC 3'
3' GTAGTTGATGTTGAGGTTTCTGTGGGAATGTGTAGTTGTTTGGATGGTG 5'
5' GTG GGT
5' TGG GTA
5' GGG TAG

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#### Protein BLAST: search protein databases using a protein query ◀ ▶ 🕂 🖇 blast.ncbi.nlm.nih.gov/Blast.cgi?PROGRAM=blastp&BLAST\_PROGRAMS=blastp& 0 Enter Query Sequence Enter accession number(s), gi(s), or FASTA sequence(s) @ Query subrange @ >gi|4504349|ref|NP\_000509.1| hemoglobin subunit beta [Homo sapiens] MYHLTPEEKSAVTALWGKVNVDEVGGEALGRLLVVYPWTQRFFESFGDLSTPDAVMGNPKVKAHGK KVLGAFSDGLAHLDNLKGTFATLSELHCDKLHVDPENFRLLGNVLVCVLAHHFGKEFTPPVQAAYQK VVAGVANALAHKYH Or, upload file Choose File no file selected Enter a descriptive title for your BLAST search @ Alian two or more sequences ( Choose Search Set Non-redundant protein sequences (nr) 🛊 🥹 Organism Fyclude + Enter organism common name, binomial, or tax id. Only 20 top taxa will be shown. 🥝 Exclude Entrez Query Enter an Entrez query to limit search Program Selection blastp (protein-protein BLAST) PSI-BLAST (Position-Specific Iterated BLAST) OPHI-BLAST (Pattern Hit Initiated BLAST) O DELTA-BLAST (Domain Enhanced Lookup Time Accelerated BLAST) Choose a BLAST algorithm (a) Search database Non-redundant protein sequences (nr) using Blastp (protein-protein BLAST) BLAST Show results in a new window + Algorithm parameters

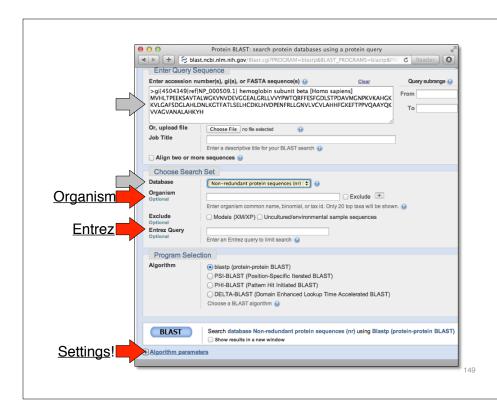
#### Step 3: Choose the database

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



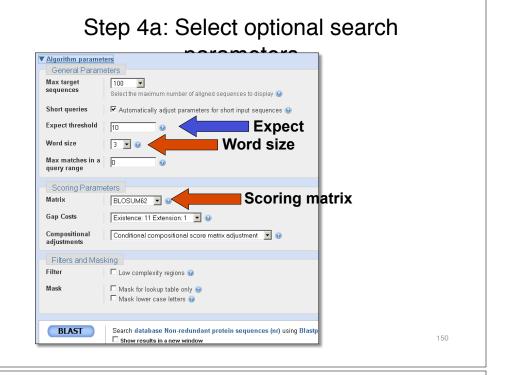
nucleotide databases



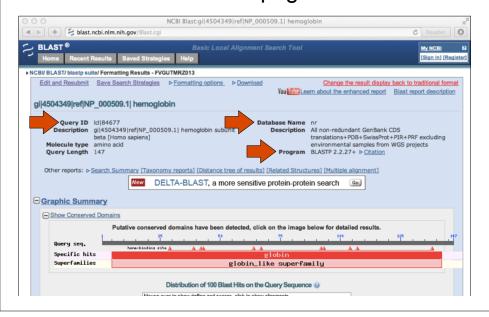


#### Step 4: Optional parameters

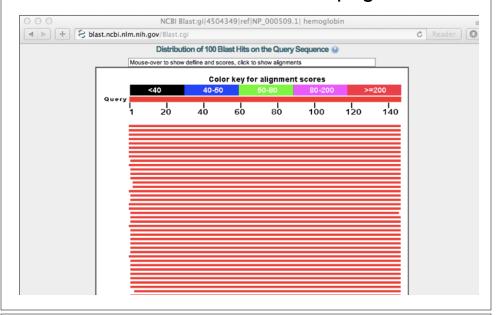
- You can...
  - choose the organism to search
  - change the substitution matrix
  - change the expect (E) value
  - change the word size
  - change the output format

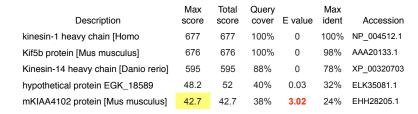


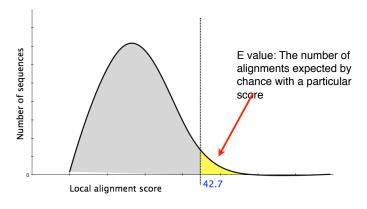
#### Results page



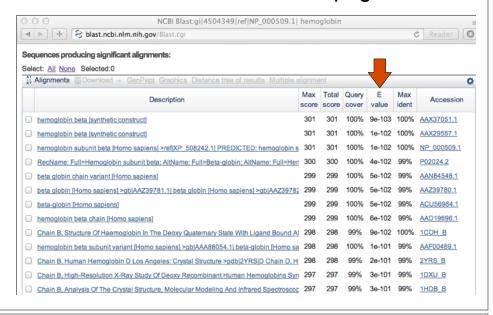
#### Further down the results page...







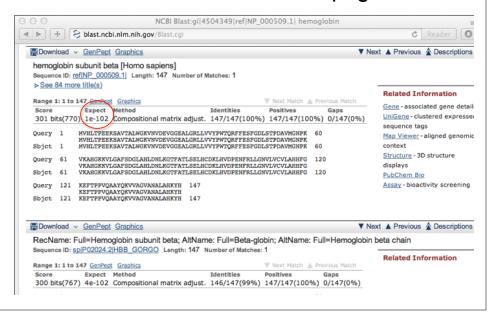
#### Further down the results page...



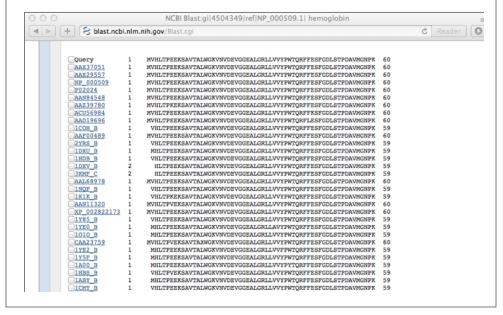
#### E values in BLAST

- Each alignment gets a score determined from the alignment and doesn't take into account the full length of the query, target or database
- The E value is what you want to look at
- E value = Expect
  - How often do I expect an alignment with this score give the length of my query and the size of the database
  - $E = Kmne^{-\lambda s}$ 
    - K and  $\lambda$  are scaling factors
    - · S is the score
    - m length of query, n length of database
  - E corrects for multiple comparisons, i.e., query compared to many sequences – proportional to length of database and query for a given S (score)

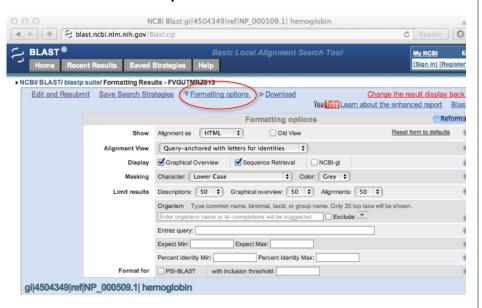
#### Further down the results page...



#### 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 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)

#### 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