# Programming for Biology Similarity Searching II –

# Practical search strategies

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# Protein Evolution and Sequence Similarity

#### **Similarity Searching I**

- · What is Homology and how do we recognize it?
- How do we measure sequence similarity alignments and scoring matrices?
- · DNA vs protein comparison

#### Similarity Searching II

- · More effective similarity searching
  - Smaller databases
  - Appropriate scoring matrices
  - Using annotation/domain information

# Similarity Searching II

- 1. What question to ask?
- 2. What program to use?
- 3. What database to search?
- 4. How to avoid mistakes (what to look out for)
- 5. When to do something different
- 6. More sensitive methods (PSI-BLAST, HMMER)

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## 1. What question to ask?

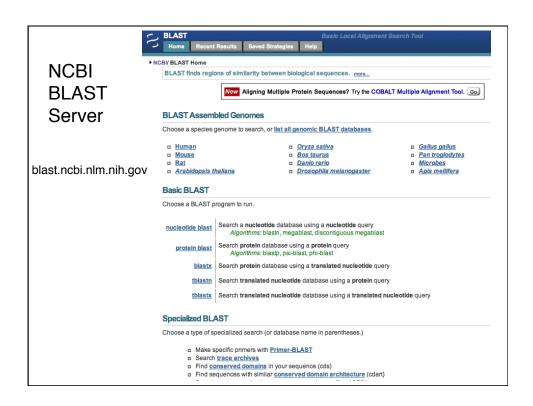
- Is there an homologous protein (a protein with a similar structure)?
- Does that homologous protein have a similar function?
- Does XXX genome have YYY (kinase, GPCR, ...)?

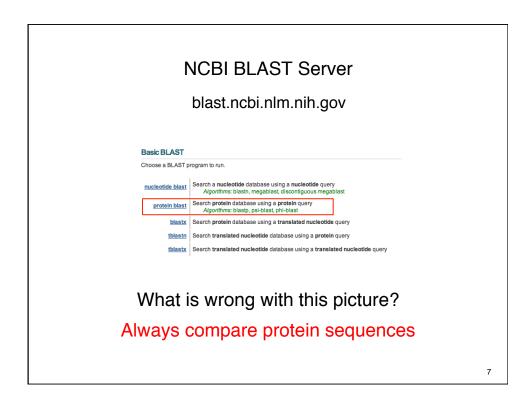
#### Questions not to ask:

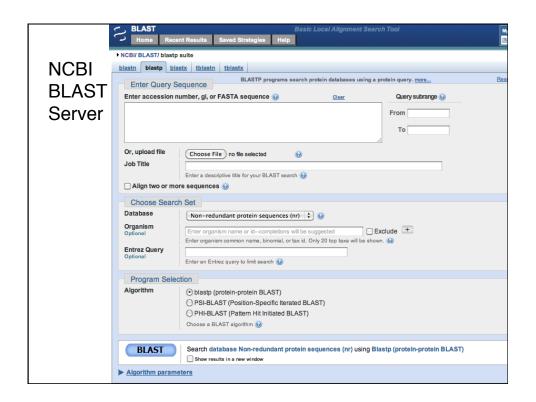
- Does this DNA sequence have a similar regulatory element (too short – never significant)?
- Does (non-significant) protein have a similar function/modification/antigenic site?

## 2. What program to run?

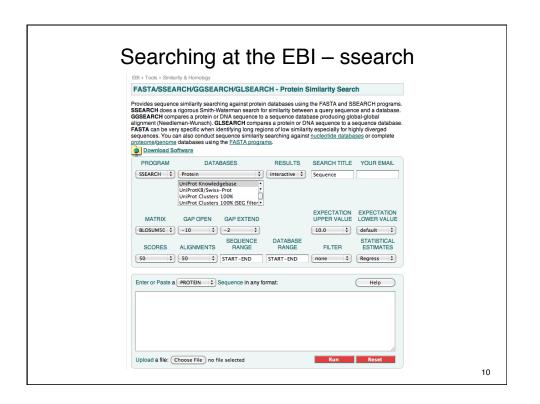
- What is your query sequence?
  - protein BLAST (NCBI), SSEARCH (EBI)
  - protein coding DNA (EST) –
     BLASTX (NCBI), FASTX (EBI)
  - DNA (structural RNA, repeat family) BLASTN (NCBI), FASTA (EBI)
- Does XXX genome have YYY (protein)?
  - TBLASTN YYY vs XXX genome
  - TFASTX YYY vs XXX genome
- Does my protein contain repeated domains?
  - LALIGN (UVa http://fasta.bioch.virginia.edu)









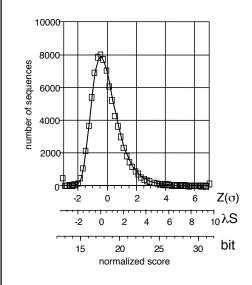


#### 3. What database to search?

- Search the smallest comprehensive database likely to contain your protein
  - vertebrates human proteins (40,000)
  - fungi S. cerevisiae (6,000)
  - bacteria E. coli, gram positive, etc. (<100,000)</li>
- Search a richly annotated protein set (SwissProt, 450,000)
- Always search NR (> 12 million) LAST
- Never Search "GenBank" (DNA)

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#### Why smaller databases are better - statistics



$$S' = \lambda S_{raw} - ln K m n$$
  
 $S_{bit} = (\lambda S_{raw} - ln K)/ln(2)$   
 $P(S'>x) = 1 - exp(-e^{-x})$   
 $P(S_{bit} > x) = 1 - exp(-mn2^{-x})$   
 $E(S'>x ID) = P D$ 

#### What is a "bit" score?

- Scoring matrices (PAM250, BLOSUM62, VTML40) contain "log-odds" scores:

  - $s_{i,j}$  (bits) =  $log_2(q_{i,j}/p_ip_j)$  ( $q_{i,j}$  freq. in homologs/  $p_ip_j$  freq. by chance)  $s_{i,j}$  (bits) = 2 -> a residue is  $2^2$ =4-times more likely to occur by homology compared with chance (at one residue)
  - $s_{ij}$  (bits) = -1 -> a residue is  $2^{-1}$  = 1/2 as likely to occur by homology compared with chance (at one residue)
- An alignment score is the maximum sum of  $s_{i,j}$  bit scores across the aligned residues. A 40-bit score is  $2^{40}$  more likely to occur by homology than by chance.
- How often should a score occur by chance? In a 400 \* 400 alignment, there are ~160,000 places where the alignment could start by chance, so we expect a score of 40 bits would occur:  $P(S_{bit} > x) = 1 - exp(-mn2^x) \sim mn2^x + 400 \times 400 \times 2^{-40} = 1.6 \times 10^5 / 2^{40} (10^{13.3}) = 1.5 \times 10^{-7}$  times

Thus, the probability of a 40 bit score in ONE alignment is  $\sim 10^{-7}$ But we did not ONE alignment, we did 4,000, 40,000, 400,000, or 16 million alignments when we searched the database:

```
E(S_{bit} \mid D) = p(40 \text{ bits}) \times database size}
E(40 \mid 4,000) = 10^{-7} \times 4,000 = 4 \times 10^{-4}
                                                                            (significant)
E(40 \mid 40,000) = 10^{-7} \times 4 \times 10^{4} = 4 \times 10^{-3}
                                                                            (not significant)
E(40 \mid 400,000) = 10^{-7} \times 4 \times 10^{5} = 4 \times 10^{-2}
                                                                            (not significant)
E(40 \mid 16 \text{ million}) = 10^{-7} \text{ x } 1.6 \text{ x } 10^{7} = 1.6
                                                                            (not significant)
```

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# How many "bits" do I need?

```
E(p \mid D) = p(40 \text{ bits}) \times \text{database size}
```

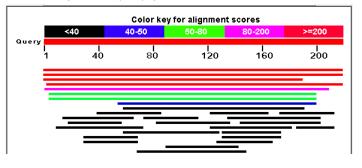
 $E(40 \mid 4,000) = 10^{-8} \times 4,000 = 4 \times 10^{-5}$ (significant)

 $E(40 \mid 40.000) = 10^{-8} \times 4 \times 10^{4} = 4 \times 10^{-4}$ (significant)

 $E(40 \mid 400,000) = 10^{-8} \times 4 \times 10^{5} = 4 \times 10^{-3}$  (not significant)

To get E()  $\sim 10^{-3}$  :

genome (10,000) p ~  $10^{-3}/10^4 = 10^{-7}/160,000 = 40$  bits SwissProt (500,000) p ~  $10^{-3}/10^6 = 10^{-9}/160,000 = 47$  bits Uniprot/NR (10<sup>7</sup>)  $p \sim 10^{-3}/10^7 = 10^{-10}/160,000 = 50$  bits



very significant 10<sup>-50</sup> significant 10<sup>-6</sup>

significant 10-3 not significant

# E()-values when??

- E()-values (BLAST expect) provide accurate statistical estimates of similarity by chance
  - non-random -> not unrelated (homologous)
  - E()-values are accurate (0.001 happens 1/1000 by chance)
  - E()-values factor in (and depend on) sequence lengths and database size
- E()-values are NOT a good proxy for evolutionary distance
  - doubling the length/score SQUARES the E()-value
  - percent identity (corrected) reflects distance (given homology)

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#### NCBI - selecting sequences with Entrez NCBI/ BLAST/ blastp suite <u>blastn</u> blastp <u>blastx</u> <u>tblastn</u> <u>tblastx</u> BLASTP programs search protein databases using a protein query. more.. Enter Query Sequence Enter accession number, gi, or FASTA sequence @ Query subrange (2) То Or, upload file Choose File no file selected Job Title Align two or more sequences (9) Choose Search Set Database Reference proteins (refseq\_protein) Organism ☐ Exclude + **Entrez Query** Enter an Entrez query to limit search (2) 16

## **Effective Similarity Searching**

- 1. Always search protein databases (possibly with translated DNA)
- Use E()-values, not percent identity, to infer homologyE() < 0.001 is significant in a single search</li>
- 3. Search smaller (comprehensive) databases
- 4. Change the scoring matrix for:
  - short sequences (exons, reads)
  - short evolutionary distances (mammals, vertebrates, aproteobacteria)
  - high identity (>50% alignments) to reduce over-extension
- 5. All methods (pairwise, HMM, PSSM) miss homologs, and find homologs the other methods miss

## Scoring matrices

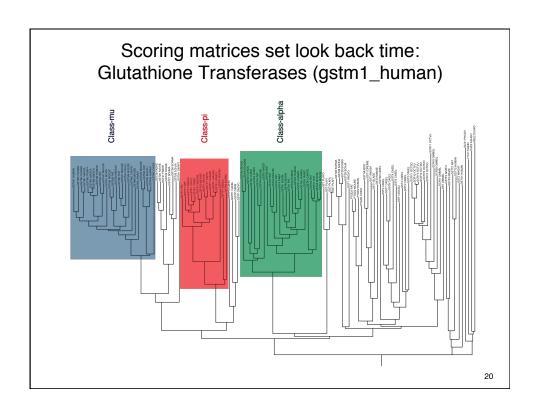
- Scoring matrices can set the evolutionary lookback time for a search
  - Lower PAM (PAM10/VT10 ... PAM/VT40) for closer (10% ... 50% identity)
  - Higher BLOSUM for higher conservation (BLOSUM50 distant, BLOSUM80 conserved)
- Shallow scoring matrices for short domains/short queries (metagenomics)
  - Matrices have "bits/position" (score/position), 40 aa at 0.45 bits/position (BLOSUM62) means 18 bit ave. score (50 bits significant)
- Deep scoring matrices allow alignments to continue, possibly outside the homologous region

# Where do scoring matrices come from?

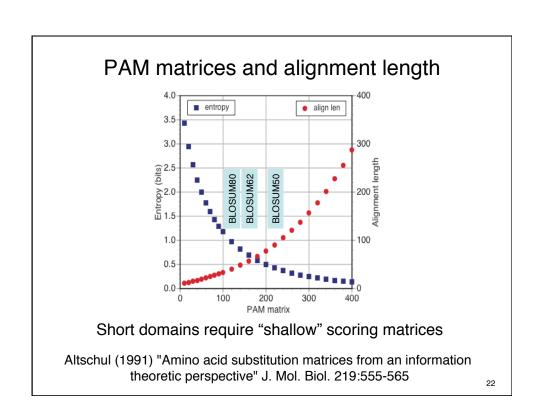
F	Pam	40						Р	am	250					
	Α	R	N	D	E	I	L		Α	R	N	D	Е	I	L
I	8							A	2						
F	R <b>-</b> 9	12						R	-2	6					
1	<b>1</b> –4	<b>-</b> 7	11					N	0	0	2				
Ι	-4	-13	3	11				D	0	-1	2	4			
Ε	-3	-11	-2	4	11			E	0	-1	1	3	4		
]	-6	-7	-7	-10	<b>-</b> 7	12		I	-1	-2	-2	-2	-2	5	
Ι	-8	-11	<b>-</b> 9	-16	-12	-1	10	L	-2	-3	-3	-4	-3	2	6

$$\lambda S_{i,j} = \log_b(\frac{q_{i,j}}{p_i p_j})$$

 $\begin{array}{ll} q_{ij} : \text{replacement frequency at PAM40, } 250 \\ q_{R:N~(~40)} = 0.000435 & p_R = 0.051 \\ q_{R:N~(250)} = 0.002193 & p_N = 0.043 \\ \textbf{l}_2 ~S_{ij} = \textbf{lg}_2 ~(q_{ij}/p_ip_j) & \textbf{l}_e ~S_{ij} = \textbf{ln}(q_{ij}/p_ip_j) & p_Rp_N = 0.002193 \\ \textbf{l}_2 ~S_{R:N(~40)} = \textbf{lg}_2 ~(0.000435/0.00219) = -2.333 \\ \textbf{l}_2 = 1/3; ~S_{R:N(~40)} = -2.333/\textbf{l}_2 = -7 \\ \textbf{l} ~S_{R:N(250)} = \textbf{lg2} ~(0.002193/0.002193) = ~0 \end{array}$ 



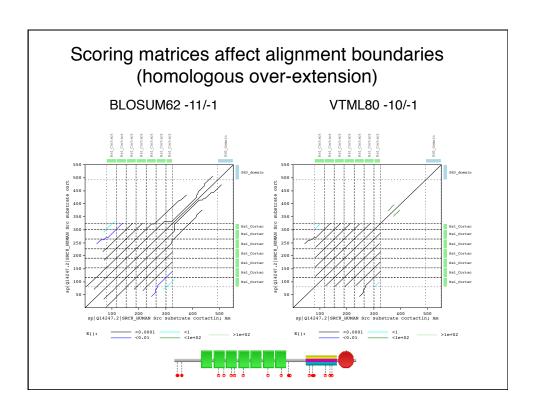
	1	BLOSUM50-10/-2	BLOSUM62-	11/-1	<b>ህ</b> ጥ40	-21/-4	VT10 -23/-4
							E(320363) f_id
				_			
	GSTM1_HUMAN	1.3e-101 1.00				1.000	0 1.000
	GSTM4_HUMAN	1.9e-89 0.867					1.9e-193 0.867
	GSTM2_MOUSE GSTM5 HUMAN	3.0e-87 0.839					2.5e-187 0.847 7.2e-195 0.912
Class-mu	GSTM3_HUMAN	8.2e-87 0.844					1.3e-184 0.844
Class-IIIu	GSTM1 MOUSE	7.0e-83 0.780					
	GSTM6 MOUSE						1.3e-161 0.779
	GSTM4 MOUSE						2.1e-158 0.769
	GSTM5_MOUSE	6.9e-73 0.727					3.7e-128 0.727
	GSTM3_HUMAN	8.2e-73 0.731	6.7e-95 (	.731	3.4e-143	0.731	8.2e-129 0.731
	GSTM2 CHICK	9.8e-65 0.656	4 70 94 (	656	2 00 117	0 656	1.4e-93 0.675
	GSTM2_CHICK GST26 FASHE	2.9e-44 0.495					
	GSTM1 DERPT	5.2e-42 0.467					
	GST27 SCHMA	2.4e-37 0.467					5.1e-20 0.607
	_						
	GSTP1_PIG	2.9e-20 0.327				0.409	
Class-pi	GSTP1_XENLA	5.2e-19 0.333				0.464	
оо	GSTP2_MOUSE GSTP1 CAEEL	8.0e-17 0.294 1.1e-16 0.324				0.395 0.706	
	GSTP1_CAEEL GSTP1 HUMAN	3.0e-16 0.284				0.467	
	GSTP1 BUFBU	1.2e-14 0.285			9.7	0.588	
	GSTPA_CAEEL	1.1e-13 0.298			0.002	0.400	
	pmana wayan	4 0 10 0 202	2 6: 14 6				
	PTGD2_MOUSE PTGD2 RAT	4.8e-12 0.302 4.8e-12 0.302					
		1.1e-11 0.292					
	PTGD2_HOLLK	9.8e-11 0.304					
	GSTP2 BUFBU	2.0e-10 0.288	2.2e-12 (	.307			
	GST_MUSDO	5.8e-09 0.257	2.3e-11 (	.251			
	GST1_DROME	1.0e-08 0.255	2.9e-10 (	.237			
	GSTA1 MOUSE	1.5e-08 0.279	4 9e-11 (	264			
	GSTA2 HUMAN	6.6e-08 0.286					
Class-	GSTA5 HUMAN	7.8e-08 0.275					
alpha	GSTA2 MOUSE	1.1e-07 0.269					
•	GSTA3_MOUSE	1.3e-07 0.278	8.9e-09 (	.258			
	GSTA1_HUMAN	3.0e-07 0.272	8.0e-08 (				
	GST36_CAEEL	3.3e-07 0.256	1.1e-08 (				
	GSTA2_CHICK	4.2e-07 0.279	8.0e-08 (	266			



# Empirical matrix performance (median results from random alignments)

Matrix			
VT160 -12/-2	23.8	0.26	192
BLOSUM50 -10/-2	25.3	0.23	217
BLOSUM62* -11/-1	28.9	0.45	111
VT120 -11/-1	27.4	1.03	48
VT80 -11/-1	51.9	1.55	32
PAM70* -10/-1	33.8	0.64	78
PAM30* -9/-1	45.5	1.06	47
VT40 -12/-1	72.7	2.76	18
VT20 -15/-2	84.6	3.62	13
VT10 /16/-2	90.9	4.32	12

HMMs can be very "deep"



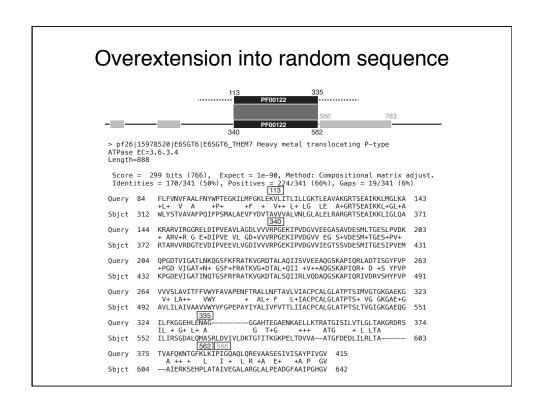
# Scoring Matrices - Summary

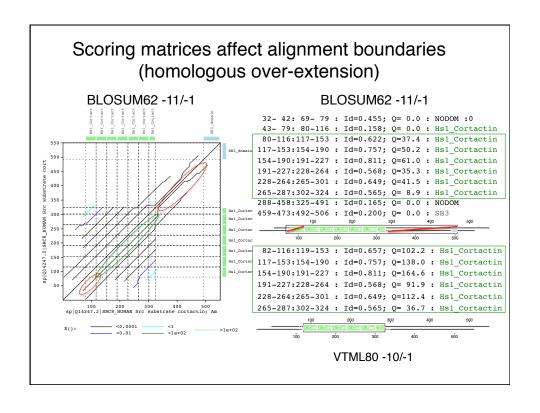
- PAM and BLOSUM matrices greatly improve the sensitivity of protein sequence comparison – low identity with significant similarity
- PAM matrices have an evolutionary model lower number, less divergence – lower=closer; higher=more distant
- BLOSUM matrices are sampled from conserved regions at different average identity – higher=more conservation
- · Shallow matrices set maximum look-back time
- Short alignments (domains, exons, reads) require shallow (higher information content) matrices

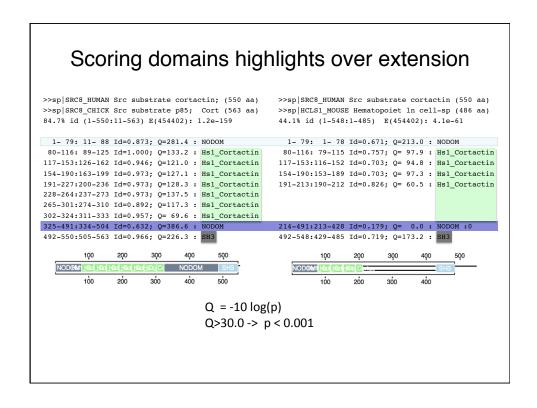
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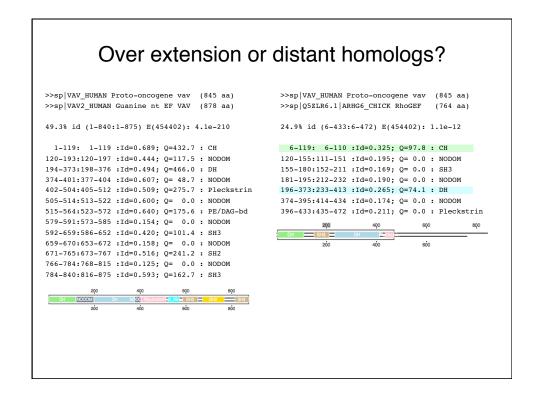
#### Effective Similarity Searching Using Annotations

- Modern sequence similarity searching is highly efficient, sensitive, and reliable – homologs are homologs
  - similarity statistics are accurate
  - databases are large
  - most queries will find a significant match
- Improving similarity searches
  - smaller databases
  - appropriate scoring matrices for short reads/assemblies
  - appropriate alignment boundaries
- Extracting more information from annotations
  - homologous over extension
  - scoring sub-alignments to identify homologous domains
- All methods (pairwise, HMM, PSSM) miss homologs
  - all methods find genuine homologs the other methods miss









#### Homology, non-homology, and over-extension

- Sequences that share statistically significant sequence similarity are homologous (simplest explanation)
- But not all regions of the alignment contribute uniformly to the score
  - lower identity/Q-value because of non-homology (overextension) ?
  - lower identity/Q-value because more distant relationship (domains have different ages) ?
- · Test by searching with isolated region
  - can the <u>distant domain (?)</u> find closer (significant) homologs?
- Similar (homology) or distinct (non-homology) structure is the gold standard
- · Multiple sequence alignment can obscure over-extension
  - if the alignment is over-extended, part of the alignment is NOT homologous

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   E() < 0.001 is significant in a single search</li>
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### **Effective Similarity Searching Using Annotations**

- Use protein/translated DNA comparisons
- Modern sequence similarity searching is highly efficient, sensitive, and reliable – homologs are homologs
  - similarity statistics are accurate
  - databases are large
  - most queries will find a significant match
- Improving similarity searches
  - smaller databases
  - shallow scoring matrices for short reads/assemblies
  - shallow matrices for high identity alignments
- Extracting more information from annotations
  - homologous over extension
  - scoring sub-alignments to identify homologous domains
- · All methods (pairwise, HMM, PSSM) miss homologs
  - all methods find genuine homologs the other methods miss

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- 1. Always search protein databases (possibly with translated DNA)
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