Reminders!

- PLEASE Register for the Virtual Bioinformatics Conference It's Free.
- September 24-26, 2002, Access Grid, Room ECS 212.

- Homework #1
- Run the BLAST Tutorial

Useful Terms

- E value: Expected # of chance alignments with scores ≥ S. The lower the E value, the more significant the score.
- P value: The probability of an alignment occurring with score ≥ S for a random sequence. Calculated by relating the observed score, S, to the expected distribution of HSP scores from comparisons of random sequences of the same length and composition as the query to the database. The most highly significant P values will be those close to 0.
- **HSP:** High-scoring segment pair. Local alignments with no gaps that achieve high alignment scores

Rules of Thumb

- Most sequences with significant similarity over their entire lengths are homologous.
- Matches that are > 50% identical in a 20-40 aa region occur frequently by chance.
- Distantly related homologs may lack significant similarity. Homologous sequences may have few absolutely conserved residues.
- A homologous to B & B to $C \Rightarrow A$ homologous to C.
- Low complexity regions, transmembrane regions and coiled-coil regions frequently display significant similarity without homology.
- Greater evolutionary distance implies that length of a local alignment required to achieve a statistically significant score also increases.

Rules of Thumb

- Results of searches using different scoring systems may be compared directly using normalized scores.
- If S is the (raw) score for a local alignment, the normalized score S' (in bits) is given by

 $S' = \frac{\lambda - \ln(K)}{\ln(2)}$

The parameters depend on the scoring system.

Statistically significant normalized score,

$$S' > \log\!\left(\frac{N}{E}\right)$$

where E-value = E, and N = size of search space.

Homologs: Orthologs & Paralogs

- **Homology:** Similarity due to common ancestry.
- Orthologs: Homologous sequences in different species that arose from a common ancestral gene during speciation; may or may not be responsible for a similar function.
- **Paralogs:** Homologous sequences within a single species that arose by gene duplication.



Growth of SWISS-PROT



09/24/2002

Number of entries

Lecture 8

Amino-acid composition from SWISS-PROT

Amino acid composition



Lecture 8

PDB Growth



09/24/2002

Lecture 8

Growth in New Folds - PDB



Multiple Alignments

- Family alignment for the ITAM domain
- CD3D_MOUSE/1-2
 Q90768/1-21
 DQLYQPLGER NDGQ-YSQLA TA
 CD3G_SHEEP/1-2
 DQLYQPLGER NDGQ-YSQLA TA
 DQ3G_SHEEP/1-2
 DQLYQPLKER EDDQ-YSHLR KK
 P79951/1-21
 NDLYQPLGQR SEDT-YSHLN SR
 FCEG_CAVPO/1-2
 DGIYTGLSTR NQET-YETLK HE
 CD3Z_HUMAN/3-0
 DGLYQGLSTA TKDT-YDALH MQ
 C79A_BOVIN/1-2
 DHTYEGLNLD DCSM-YEDIS RG
 C79B_MOUSE/1-2
 DHTYEGLNID QTAT-YEDIV TL
 CD3H_MOUSE/1-2
 NPVYNELNUG RREE-YDVLE KK
 CD3Z_SHEEP/1-2
 NPVYNELNUG RREE-YAVLD RR
 CD3E_HUMAN/1-2
 NPDYEPIRKG QRDL-YSGLN QR
 CD3H_MOUSE/2-0
 CONSENSUS/60%

CLUSTALW

- * identical
- : conserved substitutions
- . semi-conserved substitutions

gi 2213819	CDN-ELKSEAIIEHLCASEFALRMKIKEVKKENGDKK	223
gi 12656123	ELKSEAIIEHLCASEFALRMKIKEVKKENGD	31
gi 7512442	CKNKNDDDNDIMETLCKNDFALKIKVKEITYINRDTK	211
gi 1344282	QDECKFDYVEVYETSSSGAFSLLGRFCGAEPPPHLVSSHHELAVLFRTDH	400

Red:AVFPMLW (Small & hydrophobic)Blue:DE (Acidic)Magenta:RHK (Basic)Green:STYHCNGQ (Hydroxyl, Amine, Basic)Gray:Others

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How to Score Multiple Alignments?

- Sum of Pairs Score (SP)
 - Optimal alignment: O(d^N) [Dynamic Prog]
 - Approximate Algorithm: Approx Ratio 2
 - Locate Center: O(d²N²)
 - Locate Consensus: O(d²N²)

Consensus char: char with min distance sum Consensus string: string of consensus char Center: input string with min distance sum

Multiple Alignment Methods

- Phylogenetic Tree Alignment (NP-Complete)
 Given tree, task is to label leaves with strings
- Iterative Method(s)
 - Build a MST using the distance function
- Clustering Methods
 - Hierarchical Clustering
 - K-Means Clustering

Multiple Alignment Methods (Cont'd)

- Gibbs Sampling Method
 - Lawrence, Altschul, Boguski, Liu, Neuwald, Winton, Science, 1993
- Hidden Markov Model
 - Krogh, Brown, Mian, Sjolander, Haussler, JMB, 1994

Profile Method

PROFILE METHOD, [M. Gribskov et al., '90]

Location		S	Sec	lue	nc	е		Protein
in Seq.	1	2	3	4	5	6	7	Name
14	G	V	S	A	S	Α	V	Ka RbtR
32	G	v	S	Е	М	т	I	Ec DeoR
33	G	v	S	Ρ	G	т	I	Ec RpoD
76	G	A	G	I	A	Т	I	Ec TrpR
178	G	С	S	R	Е	т	v	Ec CAP
205	С	L	S	Ρ	S	R	L	Ec AraC
210	C	L	S	Ρ	S	R	L	St AraC
13	G	v	Ν	K	Е	т	I	Br MerR

FREQUENCY TABLE

	A	C	D	Е	F	G	Η	I	Κ	L	М	Ν	Ρ	Q	R	S	т	V	W	Y
1	0	2	0	0	0	6	0	0	0	0	0	0	0	0	0	0	0	0	0	0
2	1	1	0	0	0	0	0	0	0	2	0	0	0	0	0	0	0	4	0	0
3	0	0	0	0	0	1	0	0	0	0	0	1	0	0	0	6	0	0	0	0
4	1	0	0	1	0	0	0	1	1	0	0	0	3	0	1	0	0	0	0	0
5	1	0	0	2	0	1	0	0	0	0	1	0	0	0	0	3	0	0	0	0
6	1	0	0	0	0	0	0	0	0	0	0	0	0	0	2	0	5	0	0	0
7	0	0	0	0	0	0	0	4	0	2	0	0	0	0	0	0	0	2	0	0

Profile Method

FREQUENCY TABLE

	Α	С	D	Е	F	G	Η	Ι	Κ	L	М	N	Ρ	Q	R	S	Т	V	W	Y
1	0	2	0	0	0	6	0	0	0	0	0	0	0	0	0	0	0	0	0	0
2	1	1	0	0	0	0	0	0	0	2	0	0	0	0	0	0	0	4	0	0
3	0	0	0	0	0	1	0	0	0	0	0	1	0	0	0	6	0	0	0	0
4	1	0	0	1	0	0	0	1	1	0	0	0	3	0	1	0	0	0	0	0
5	1	0	0	2	0	1	0	0	0	0	1	0	0	0	0	3	0	0	0	0
6	1	0	0	0	0	0	0	0	0	0	0	0	0	0	2	0	5	0	0	0
7	0	0	0	0	0	0	0	4	0	2	0	0	0	0	0	0	0	2	0	0

WEIGHT MATRIX

	A	C	E	G	I	K	L	М	N	P	R	S
1	0	108	0	101	0	0	0	0	0	0	0	0
2	21	78	0	0	0	0	44	0	0	0	0	0
3	0	0	0	23	0	0	0	0	46	0	0	102
4	21	0	32	0	38	32	0	0	0	86	39	0
5	21	0	62	23	0	0	0	74	0	0	0	72
6	21	0	0	0	0	0	0	0	0	0	69	0
7	0	0	0	0	98	0	44	0	0	0	0	0

$$Weight[i, AA] = \log\left(\frac{Freq[i, AA]}{p[AA] \cdot N}\right) \cdot 100$$

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Profile Method

WEIGHT MATRIX

	A	C	E	G	I	K	L	M	N	P	R	S
1	0	108	0	101	0	0	0	0	0	0	0	0
2	21	78	0	0	0	0	44	0	0	0	0	0
3	0	0	0	23	0	0	0	0	46	0	0	102
4	21	0	32	0	38	32	0	0	0	86	39	0
5	21	0	62	23	0	0	0	74	0	0	0	72
6	21	0	0	0	0	0	0	0	0	0	69	0
7	0	0	0	0	98	0	44	0	0	0	0	0

Given the following protein sequence:

CpG Islands

- Regions in DNA sequences with increased occurrences of substring "CG"
- Rare: typically C gets methylated and then mutated into a T.
- Often around promoter or "start" regions of genes
- Few hundred to a few thousand bases long

Problem 1:

- Input: Small sequence S
- Output: Is S from a CpG island?
 - Build Markov models: M+ and M —
 - Then compare

Markov Models

+	Α	С	G	Т		Α	С	G	Т
Α	0.180	0.274	0.426	0.120	A	0.300	0.205	0.285	0.210
С	0.171	0.368	0.274	0.188	С	0.322	0.298	0.078	0.302
G	0.161	0.339	0.375	0.125	G	0.248	0.246	0.298	0.208
Τ	0.079	0.355	0.384	0.182	Т	0.177	0.239	0.292	0.292

How to distinguish?

• Compute

$$S(x) = \log\left(\frac{P(x \mid M +)}{P(x \mid M -)}\right) = \sum_{i=1}^{L} \log\left(\frac{p_{x(i-1)x_i}}{m_{x(i-1)x_i}}\right) = \sum_{i=1}^{L} r_{x(i-1)x_i}$$

r	Α	С	G	Τ
Α	-0.740	0.419	0.580	-0.803
С	-0.913	0.302	1.812	-0.685
G	-0.624	0.461	0.331	-0.730
Τ	-1.169	0.573	0.393	-0.679

Problem 1:

- Input: Small sequence S
- Output: Is S from a CpG island?
 - Build Markov Models: M+ & M-
 - Then compare

Problem 2:

- Input: Long sequence S
- Output: Identify the CpG islands in S.
 - Markov models are inadequate.
 - Need Hidden Markov Models.