CAP 5510: Introduction to Bioinformatics

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HMM for Sequence Alignment

A. Sequence alignment

N		F	L	S
Ν	٠	F	L	S
Ν	K	Y	L	Т
Q	٠	W	-	Т

RED POSITION REPRESENTS ALIGNMENT IN COLUMN GREEN POSITION REPRESENTS INSERT IN COLUMN PURPLE POSITION REPRESENTS DELETE IN COLUMN

B. Hidden Markov model for sequence alignment



FIGURE 5.16. Relationship between the sequence alignment and the hidden Markov model of the alignment (Krogh et al. 1994). This particular form for the HMM was chosen to represent the sequence, structural, and functional variation expected in proteins. The model accommodates the identities, mismatches, insertions, and deletions expected in a group of related proteins. (*A*) A section of an msa. The illustration shows the columns generated in an msa. Each column may include matches and mismatches (*red* positions), insertions (*green* positions), and deletions (*purple* positions). (*B*) The HMM. Each column in the model represents the possibility of a match, insert, or delete in each column of the alignment in *A*. The HMM is a probabilistic representation of a section of the msa. Sequences can be generated from the HMM by starting at the beginning state labeled BEG and then by following

Problem 3: LIKELIHOOD QUESTION

- Input: Sequence S, model M, state i
- Output: Compute the probability of reaching state i with sequence S using model M
 - Backward Algorithm (DP)

Problem 4: LIKELIHOOD QUESTION

- Input: Sequence S, model M
- Output: Compute the probability that S was emitted by model M
 - Forward Algorithm (DP)

Problem 5: LEARNING QUESTION

- Input: model structure M, Training Sequence S
- Output: Compute the parameters Θ
- Criteria: ML criterion
 - maximize $P(S | M, \Theta)$ HOW???

Problem 6: DESIGN QUESTION

- Input: Training Sequence S
- Output: Choose model structure M, and compute the parameters ⊖
 - No reasonable solution
 - Standard models to pick from

Iterative Solution to the LEARNING QUESTION (Problem 5)

\Box Pick initial values for parameters Θ_0

Repeat

Run training set S on model M Count # of times transition $i \Rightarrow j$ is made Count # of times letter x is emitted from state i Update parameters Θ

Until (some stopping condition)

Entropy measures the variability observed in given data.

$$E = -\sum_{c} p_c \log p_c$$

Entropy is useful in multiple alignments & profiles.

Entropy is max when uncertainty is max.

G-Protein Couple Receptors

- \square Transmembrane proteins with 7 α -helices and 6 loops; many subfamilies
- θ Highly variable: 200-1200 as in length, some have only 20% identity.
- θ [Baldi & Chauvin, '94] HMM for GPCRs
- HMM constructed with 430 match states (avg length of sequences);
 Training: with 142 sequences, 12 iterations

GPCR - Analysis

Compute main state entropy values $H_i = -\sum_{a}^{a} e_{ia} \log e_{ia}$

□ For every sequence from test set (142) & random set (1600) & all SWISS-PROT proteins

• Compute the negative log of probability of the most probable for the $P(\pi \mid S, M)$

GPCR Analysis



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Entropy





GPCR Analysis (Cont'd)



Figure 8.2: Scores (Negative Log-likelihoods of Optimal Viterbi Paths). Represented sequences consist of 142 GPCR training sequences, all sequences from the SWISS-PROT database of length less than or equal to 2000, and 220 randomly generated sequences with same average composition as the GPCRs of length 300, 350, 400, 450, 500, 550, 600, 650, 700, 750, 800 (20 at each length). The regression line was obtained from the 220 random sequences. The horizonta' distances in the histogram correspond to (malized scores (6).

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Applications of HMM for GPCR

- Bacteriorhodopsin
 - Transmembrane protein with 7 domains
 - But it is not a GPCR
 - Compute score and discover that it is close to the regression line. Hence not a GPCR.
- Thyrotropin receptor precursors
 - All have long initial loop on INSERT STATE 20.
 - Also clustering possible based on distance to regression line.

HMMs – Advantages

- Sound statistical foundations
- Efficient learning algorithms
- Consistent treatment for insert/delete penalties for alignments in the form of locally learnable probabilities
- Capable of handling inputs of variable length
- Can be built in a modular & hierarchical fashion; can be combined into libraries.
- Wide variety of applications: Multiple Alignment, Data mining & classification, Structural Analysis, Pattern discovery, Gene prediction.

HMMs – Disadvantages

□ Large # of parameters.

Cannot express dependencies & correlations between hidden states.

Protein Structures

- Sequences of amino acid residues
- 20 different amino acids



Proteins



More on Secondary Structures

θ α -helix

- Main chain with peptide bonds
- Side chains project outward from helix
- Stability provided by H-bonds between CO and NH groups of residues 4 locations away.

θ β -strand

Stability provided by H-bonds with one or more β-strands, forming β-sheets.
 Needs a β-turn.

Proteins

Tertiary structures are formed by packing secondary structural elements into a globular structure.



Quaternary Structures in Proteins

 The final structure may contain more than one "chain" arranged in a quaternary structure.





Insulin Hexamer

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Amino Acid Types

🔲 Hydrophobic	I,L,M,V,A,F,P		
Charged			
Basic	K,H,R		
Acidic	E,D		
🖵 Polar	S,T,Y,H,C,N,Q,W		
🔲 Small	A,S,T		
🖵 Very Small	A,G		
🖵 Aromatic	F,Y,W		

Structure of a single amino acid



Chains of amino acids



Amino acids vs Amino acid residues

Angles ϕ and ψ in the polypeptide chain







Amino Acid Structures from Klug & Cummings



3. Polar: positively charged (basic)





Amino Acid Structures from Klug & Cummings



(c) David Gilbert, Aik Choon Tan, Gilleain Torrance and Mallika Veeramalai 2002 16



Alpha Helix



Beta sheet



(c) David Gilbert, Aik Choon Tan, Gilleain Torrance and Mallika Veeramalai 2002 17

Beta Strand



Active Sites

Active sites in proteins are usually hydrophobic pockets/crevices/troughs that involve sidechain atoms.



Figure 4.13 (a) The active site in open twisted α/β domains is in a crevice outside the carboxy ends of the β strands. This crevice is formed by two adjacent loop regions that connect the two strands with α helices on opposite sides of the β sheet. This is illustrated by the curled fingers of two hands (b), where the top halves of the fingers represent loop regions and the bottom halves represent the β strands. The rod represents a bound molecule in the binding crevice.





Left PDB 3RTD (streptavidin) and the first site located by the MOE Site Finder. Middle 3RTD with complexed ligand (biotin). Right Biotin ligand overlaid with calculated alpha spheres of the first site.

Secondary Structure Prediction Software



Figure 11.3 Comparison of secondary structure predictions by various methods. The sequence of flavodoxin, an α/β protein, was used as the query and is shown on the first line of the alignment. For each prediction, H denotes an α helix, E a β strand, T a β turn; all other positions are assumed to be random coil. Correctly assigned residues ture assignment given in the PDB file for flavodoxin (10FV, Smith et al., 1983).

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R

PDB: Protein Data Bank

- Database of protein tertiary and quaternary structures and protein complexes. http://www.rcsb.org/pdb/
- Over 29,000 structures as of Feb 1, 2005.
- Structures determined by
 - NMR Spectroscopy
 - X-ray crystallography
 - Computational prediction methods
- Sample PDB file: Click here [_]

Protein Folding





Modular Nature of Protein Structures



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Protein Structures

- Most proteins have a hydrophobic core.
- Within the core, specific interactions take place between amino acid side chains.
- Can an amino acid be replaced by some other amino acid?
 - Limited by space and available contacts with nearby amino acids
- Outside the core, proteins are composed of loops and structural elements in contact with water, solvent, other proteins and other structures.

Viewing Protein Structures

- SPDBV
- RASMOL
- CHIME