

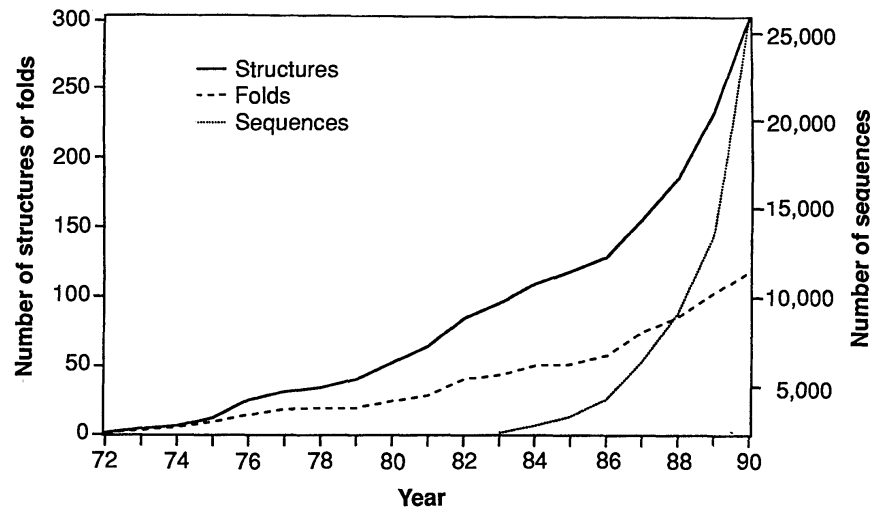
CMSC 858P Lecture 26: Protein folding – threading 5/6/08

Glossary

- Residue – any single amino-acid
- Side-chain – chemical group off the backbone
- Peptide – a short chunk of protein
- Polypeptide – protein

Threading: reverse structure prediction

- Main hypothesis: while there are many protein sequences, there are much fewer folds. I.e. nature keeps reinventing useful structures



- Given a database of structures and a query string, find which structure “fits” the string best

Initial idea: 3D-1D scores

- From a 3D structure, determine “environment” for every amino-acid
 - buried (inside the protein)
 - outside
 - inner side of helix
 - outer side of helix
 - etc...
- Annotate each position in protein with the environment information
ACKCAHGT -> $E_1 E_2 E_1 E_3 E_4 E_2 E_3 E_1 E_4$
- Why this is reasonable? Amino-acids have “preference” for specific environments

Alignment to an environment string

- Idea: use gapped alignment algorithm to estimate how likely it is for a sequence to conform to a structure (represented as an environment string)

- $$\begin{array}{cccccccccccc} E_1 E_2 - & E_1 E_3 - & - & E_4 E_2 - & E_3 E_1 E_4 \\ A & G & H & - & K & T & G & A & L & K & M & N & G \end{array}$$

- Question: what is the score of aligning an amino-acid to an environment?

Answer: use statistics

- For each environment – calculate likelihood (observed frequency) of all amino-acids based on known structures
- For each environment – empirical estimation of gap opening/extension penalties
- Alignment algorithm – use Gribskov's profile method: replace each environment character with the amino-acid frequency table for that environment

E_1

A 0.22

K 0.15

W 0.08

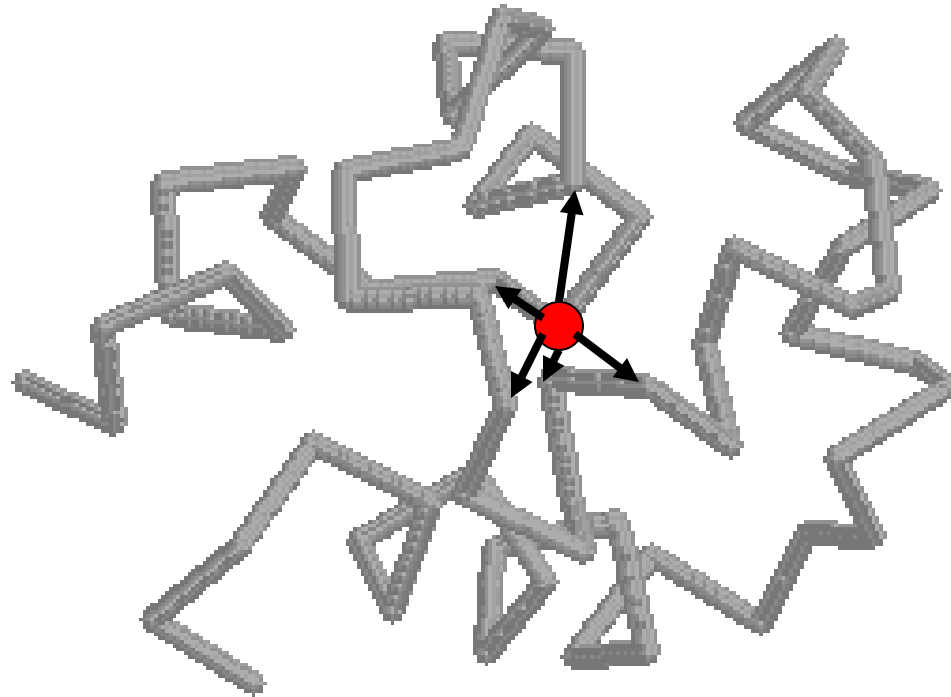
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$$S(E_1, G) = \sum_{AA} S(AA, G) * \text{freq}_{E_1}(AA)$$

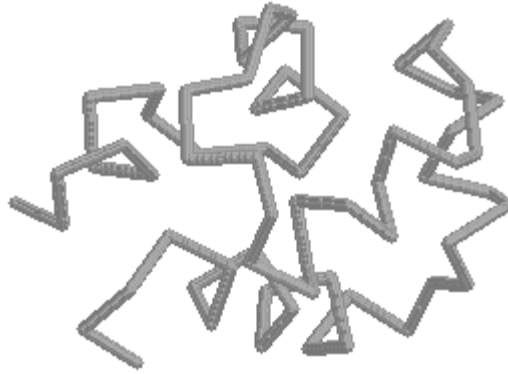
$S(AA, G)$ – e.g. from BLOSUM matrix

Environments – not good enough

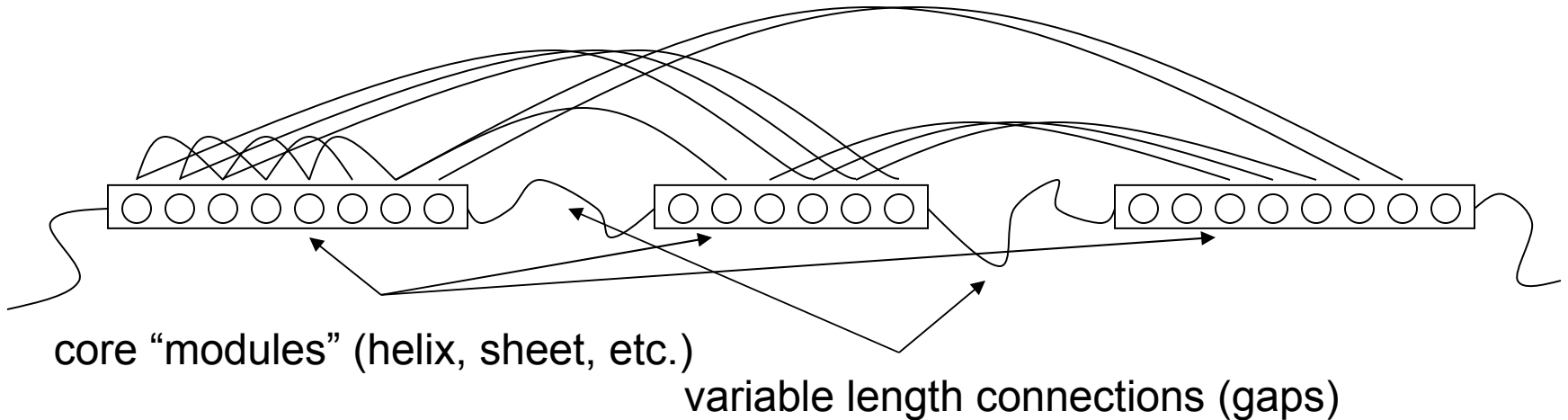
- Each amino-acid may have multiple contacts



A better model



residue interactions (and associated energy parameters)



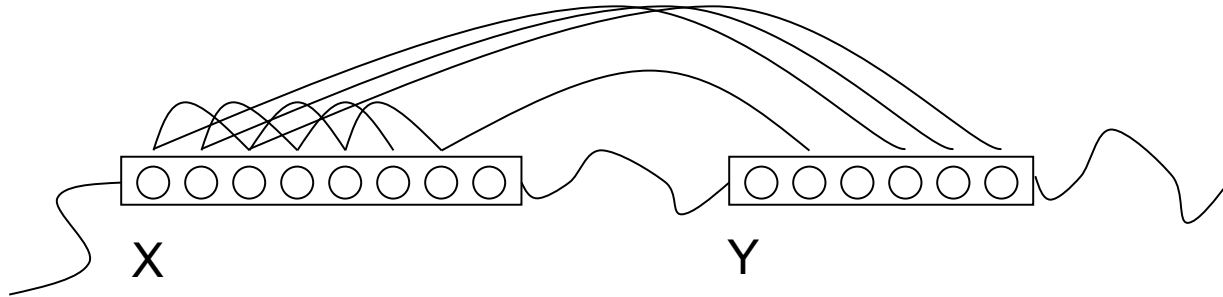
The threading problem

- Model assumptions:
 - loop AA composition and length contributes to energy score (note: can also place restrictions on minimum/maximum size in gaps)
 - interactions are pair-wise: interaction energy depends on at most two AAs
 - individual AAs in core modules also contribute to energy due to local environment
- Thread a protein sequence through a structure model s.t.
 - the place-holders are filled with amino-acids
 - a variable number of amino-acids fall in the gaps
 - overall energy is minimized
- Easy to say, hard to do: Thus defined (variable length gaps AND pair-wise interactions) the problem is NP-hard!

NP-hard => heuristics

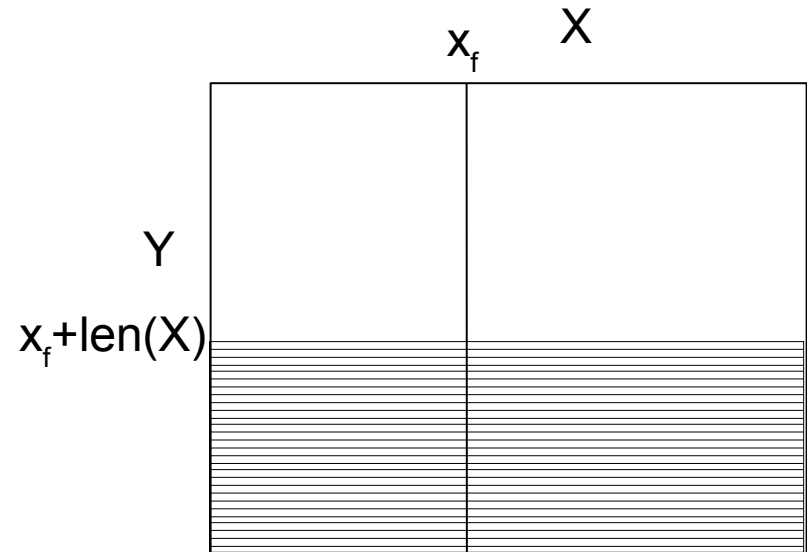
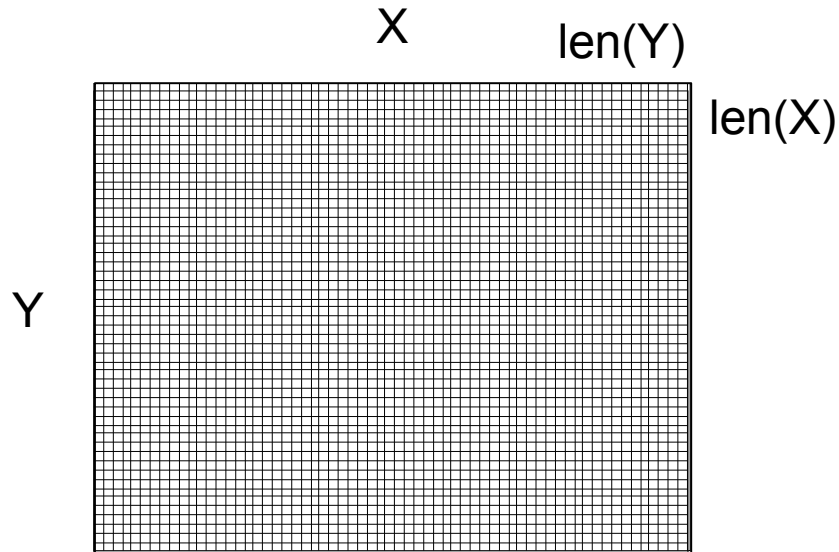
- Branch and bound (Lathrop, Smith)
 - Represent all possible folds (search space) s.t. it is easy to compute a lower bound on the score
 - Note: a threading is uniquely defined by the coordinates of the core elements – a set of threadings is a hyper-rectangle in a C-dimensional space where C is the # of core elements
 - Divide search space and compute energy lower-bounds on each sub-division (choose a dimension (core) and a coordinate and split hyper-rectangle at that location)
 - Recurse on sub-division with lowest lower-bound

Hyper-rectangle heuristic



Each “module” corresponds to a dimension - offset of module in the protein

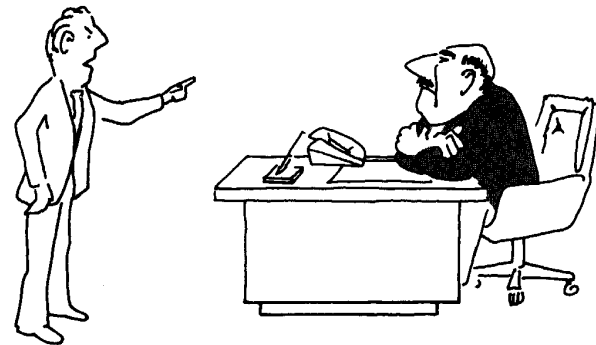
Fixing one module restricts the flexibility in assigning the remaining modules (imagine beads on a string)



NP-completeness



“I can’t find an efficient algorithm, I guess I’m just too dumb.”



“I can’t find an efficient algorithm, because no such algorithm is possible!”

From: **Computers
and**

Intractability

M. R. Garey and

D. S. Johnson

(*W. H. Freeman*

1979)



“I can’t find an efficient algorithm, but neither can all these famous people.”

Threading is NP-hard - proof

- Reduction from ONE-IN-THREE 3 SAT
 - n boolean variables, k boolean clauses with exactly 3 literals
 - 3 SAT – is there a setting of the variables such that all clauses are simultaneously true?
 - ONE-IN-THREE 3SAT – 3SAT but each clause made true by exactly one literal
- Proof: for any instance of 3SAT, create an instance of the protein threading problem s.t. a solution to the threading problem implies a solution to 3SAT

Proof ...cont

- Protein sequence
 - T, F – state of each boolean value
 - P,Q,R – which literal makes a clause true
 - protein: PQR PQR PQR...TFTFTF....
- Core model
 - one core element (with one AA) for each clause
 - one core element (with one AA) for each boolean
 - interactions from each clause to the booleans present in it. edge also encodes which literal (1,2,3) and whether value is negated
 - edge score = 0 if label consistent with amino-acid assignment and 1 otherwise (e.g. QF is consistent with edge 2,NOT)
 - optimal threading has score 0 and solves 3SAT

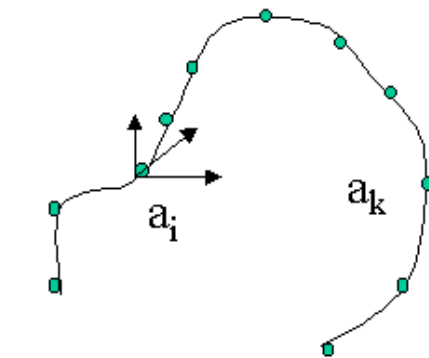
Discussion

- Both variable length gaps and pairwise interactions are essential!
- If no variable length gaps – can try all threadings in polynomial time irrespective of interactions
- If no pairwise interactions – dynamic programming can figure out the correct assignment (essentially the alignment problem)

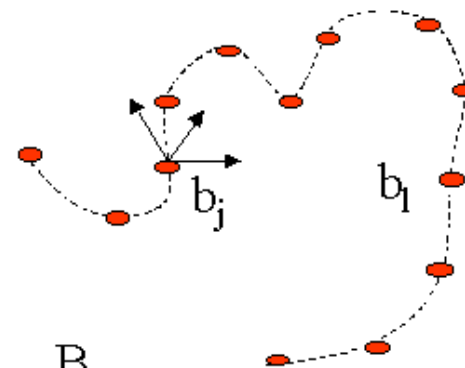
Structure to structure alignment

- Given two proteins with known structure, how do we align them to each other?
- Double Dynamic Programming
 - distance matrix depends on distance between residues
 - pick a pair of residues (i,j) and assume they are paired up
 - use dynamic programming to align the rest of the protein – score will represent score for pairing of i,j
 - use a final dynamic programming step to align the proteins based on scores determined above

Example



Structure A



B

Coordinate system
coincidence at
 a_i, b_j

