

# Protein Fold Recognition / Weak Similarities

Why do we do sequence alignments ?

- find related proteins
  - build models
  - guess at function

For some interesting protein

- sequence always available

What should one do with really weak sequence homology ?

Two ideas

- how to search for very weak similarities
- can one take advantage of conserved structures ?

# Technical

- Searching for remote sequence homologues
- Sequence to structure alignments

## Assumed knowledge

- Some memory of sequence alignment methods, score matrix,  $O(n^2)$  cost

# Mission

For some protein sequence – find as much as possible

- function
- build good model
- build a bad model

Vague information may be useful

- which residues are near active site ?
- which residues are near a dimer interface ?
- which residues are in weakly structured loops ? (chemical modification)
- bad model may be enough for phasing (X-ray)

# Approach

- start with most reliable methods
- add more speculative methods as necessary

## Example

- simple sequence searches
- searches for more remote homologues
- searches for possible structures

## Methods in other courses

- emphasis on speed (in Georgio's lectures)

# alignment methods

	slow	fast
methods	Needleman & Wunsch / Smith-Waterman	seeded – blast, fasta, suffix tree methods
time	$O(nm)$ or $O(nm^2)$ (sequence sizes)	$O(nk)$ – database size
guaranteed to find optimal alignment	yes	no
very remote homologues	may work	less likely to work

Does speed matter ?

# Slow methods

Methods for large databases are

- fast
- approximate

Here

- ultimate use is often a small database (PDB  $9.7 \times 10^4$ )
- computer time does not matter

In lab you have 1 or 10's of proteins

- each take weeks or months to work on
- if each search takes hours ? no problem

Remote searches...

# Remote searches

When to do this ?

Assume simple (blast / fasta) search returned

- related sequences
- unknown function
- none of related proteins have known structures

# Weak sequence similarities

Your sequence

**A B D E F G H I K L M N P Q...**

finds no helpful proteins. Try searching with a related protein

prot\_1      **A B Q E F G R I S L T N P Q...**

- finds a protein whose structure has been solved

prot\_2      **Q B Q E Q G R Q S L T N P A...**

Claim

- yours & prot\_2 are related
- relationship too weak to see directly
- prot\_2 can be used
  - to make a bad model, guess for function



# Weak sequence similarities

- first idea
- take your protein
- collect related proteins
  - foreach (related protein)
    - do a sequence search
    - see if results change
- not practical
- not very systematic
- what else does one get from homologues ?

# Information from related sequences

- usually one finds many related sequences.
- consider details...


```
VLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
MLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAAWGKVGAGHAGEYGAEALEKMFSLFPTTKTYFPHFDLSHGSAQVKGHG
  LSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAAWGKVGAGHAGDYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPDDKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
MLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTHVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAAWGKVGAGHAGEYGAEAWERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
MLSPADKTNVKAAWGKVGAGHAGEYGAEAWERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
MLSPADKTNVKAAYWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAHWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSAADKTNVKAGWSKVGGHAGEYGAEALERMFLGFPTTKTYFPHFDLSHGSAQVKAHG
VLSAADKTNVKAFWSKVGGHAGEYGAEALERMFLGFPTTKTYFPHFDLSHGSAQVKAHG
VLSADDKANIKAEWGKIGGHGAEYGAEALERMFCSFPTTKTYFPHFDVSHGSAQVKGHG
MLSPADKTNVKADWGKVGAGHAGEYGAEAFERMFSLFPTTKTYFPHFDLSHGSAQVKGQG
VLSPADKTNVKACWGKVGAGHAGEYGAEAFERMFSLFPTTKTYFPHFDLSHGSAQVKGQA
VLSAADKSNVKAAWGKVGGNAGAYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
MLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKSNVKATWDKIGSHAGEYGGEALERTFASFPTTKTYFPHFDLSPGSAQVKAHG
VLSPADKSNVKAWWGKVGGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
MLSPADKTNVKAAWGKVGAGHAGEYGAEALERMFSLFPTTGTYFPHFDLSHGSAQVKGHG
VLSSADKNNVKACWGKIGSHAGEYGAEALERTFCSFPTTKTYFPHFDLSHGSAQVQAHG
VLSAADKSNVKAAWGKVGGNAGAYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSPADKTNVKAQWGKVGAGHAGEYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
VLSANDKSNVKAAWGKVGNHAPEYGAEALERMFSLFPTTKTYFPHFDLSHGSSQVKAHG
VLSPADKSNVKAAWGKVGGHAGDYGAEALERMFSLFPTTKTYFPHFDLSHGSAQVKGHG
```

... ..

# Conservation

If your sequence has a Q here,

- may not be helpful to use it in sequence searches



L	D	D	Q	R	Q	S	T	R	
L	D	A	Q	R	A	D	S	T	R
V	D	D	Q	R	R	W	S	T	R
A	D	D	Q	R	C	A	S	S	K
I	D	D	Q	R	D	D	S	T	R
L	D	D	Q	R	E	G	S	T	K
L	D	D	Q	R	F	C	S	T	R

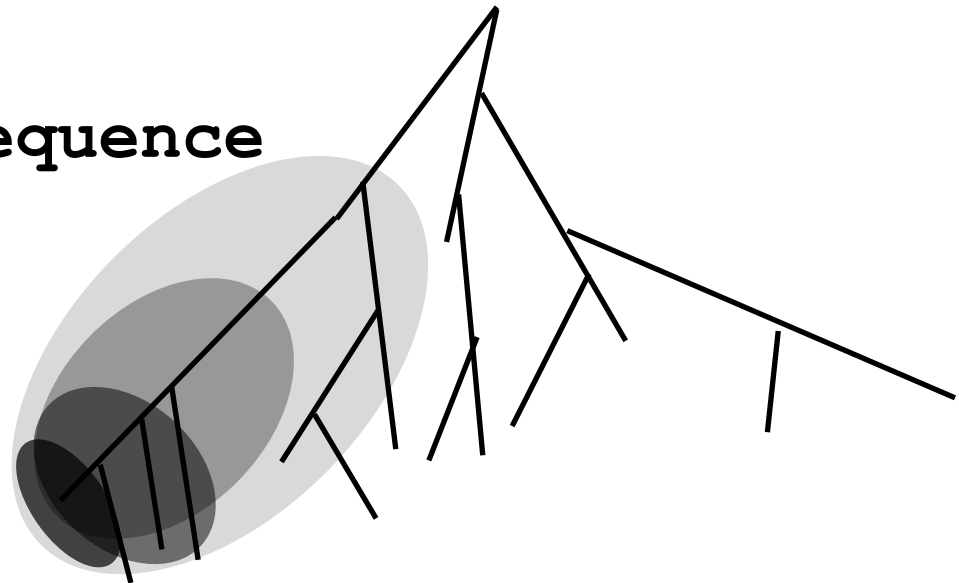
- better to use the "average" residue at this point
- first have to find the "average" residue
- leads to method

# Searching with profiles

- initial average\_sequence = your\_sequence

```
while (step < max_steps)
    search with blast using average_sequence
    if interesting result (function /
    structure..)
        return results
    else
        update average_sequence
```

- basis of "psi-blast"
- does it work ?



# Remote sequence searching

- much more sensitive than simple searches, but
- involves weaker sequence similarities, more errors
- alignment not perfect
- statistical significance harder to estimate
- possibility of finding unrelated sequences (rubbish)
- still relies on some significant sequence similarity
- can one move away from sequence similarity ?

# Why move away from sequence ?

- if sequences provide information – use this
- if you are desperate...

# Sequence alignments – implied structures

From sequence viewpoint

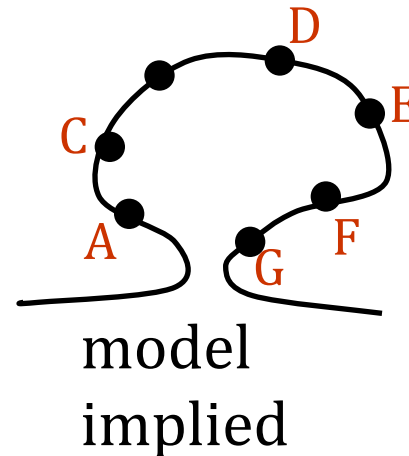
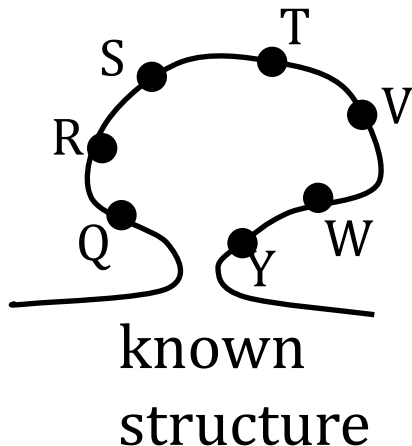
.. **AC-DEFG** ..

.. QRSTVWY ..

What if structure of second sequence is known ?

.. **AC-DEFG** .. query sequence

.. **QRSTVWY** .. known structure



# Sequence to structure alignments

Remember how sequence alignments work

- similarity / substitution scores
- fill out score matrix
- find best path

Can we use this for sequence to structure alignments ?

	A	B	C	N	J	R	O	C	L	C	R	P	M
A	4	7	6	6	5	4	4	3	3	2	1	0	0
J	7	7	6	6	6	4	4	3	3	2	1	0	0
C	6	6	7	6	5	4	4	4	3	3	1	0	0
J	6	6	6	5	6	4	4	3	3	2	1	0	0
N	5	5	5	6	5	4	4	3	3	2	1	0	0
R	4	4	4	4	4	5	4	3	3	2	2	0	0
C	3	3	4	3	3	3	3	4	3	3	1	0	0
K	3	3	3	3	3	3	3	3	3	2	1	0	0
C	2	2	3	2	2	2	2	3	2	3	1	0	0
R	2	1	1	1	1	2	1	1	1	1	2	0	0
B	1	2	1	1	1	1	1	1	1	1	1	0	0
P	0	0	0	0	0	0	0	0	0	0	0	1	0



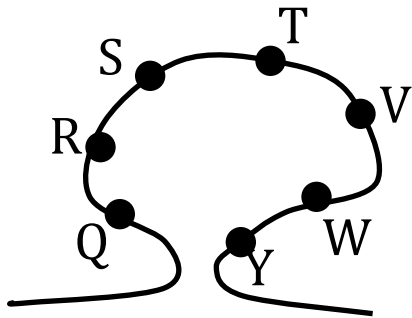
# more exotic scoring

From sequence viewpoint

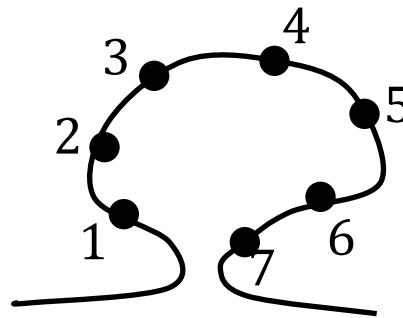
..**AC-DEFG**..      my sequence

..**QRSTVWY**..      a protein of known structure

rather than just align sequences, could I use the structure ?



known  
structure



forget  
sequence

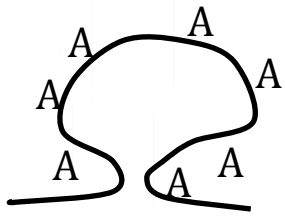
Score matrix ?

	A	C	D	E	F	G
1	?	...				
2						
3						
4						
5						
6						
7						

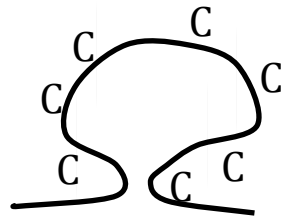
# sequence to structure scoring

I have to be able to place (A, C, D..) at each position and get a suitability score

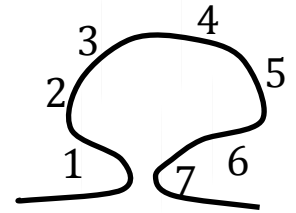
	A	C	D	E	F	G
1	?	...				
2						
3						
4						
5						
6						
7						



then



- then it would be easy to do sequence to structure alignments



Advantage:

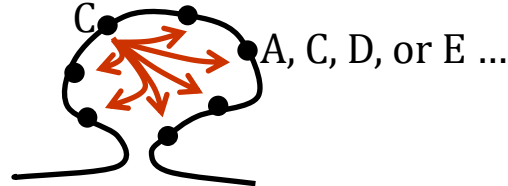
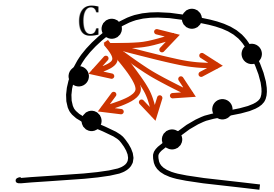
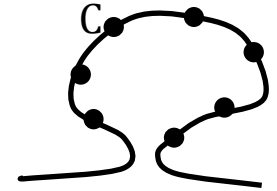
- we claim that structure is more conserved than sequence
- can find appropriate/fitting/suitable structures for a sequence
- very remote, but homologues

vorsicht !!!!

# sequence to structure scoring

Define an energy function

- depends on interaction of residue with structure
  - easy
- depends on interaction with neighbours
  - but who are the neighbours ?



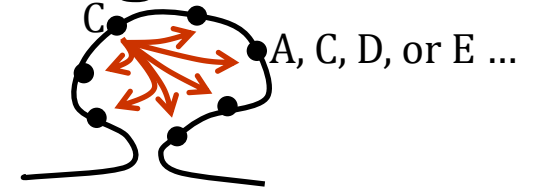
Bad news

- we cannot even fill out a column in the score matrix
- to test every combination of neighbours
  - NP-complete

An excuse to try some approximations

	A	C	D	E	F	G
1	?					
2	?					
3	?					
4	?					
5	?					
6	?					
7	?					

# approximations for scoring



## Two problems

- we do not know where all the atoms are – side chain coordinates
- to score "C" at each position we need to know neighbours

## Side-chains : ignore / average

- forget for these lectures

## Neighbour positions - much harder

- environment description
- frozen approximation

# Environment description

An example of profiles (case study)

We know

- certain sites are hidden from solvent (middle of protein)
  - only compatible with trp, phe, ile, ... (hydrophobic)
- some sites are involved in "salt bridges"
- some secondary structures are preferred by certain residues
- can one count the probabilities of residue types ?

Overview

- collect list (parameterisation set) of proteins
- classify sites (18 types)
- collect probability of each residue type in each site type

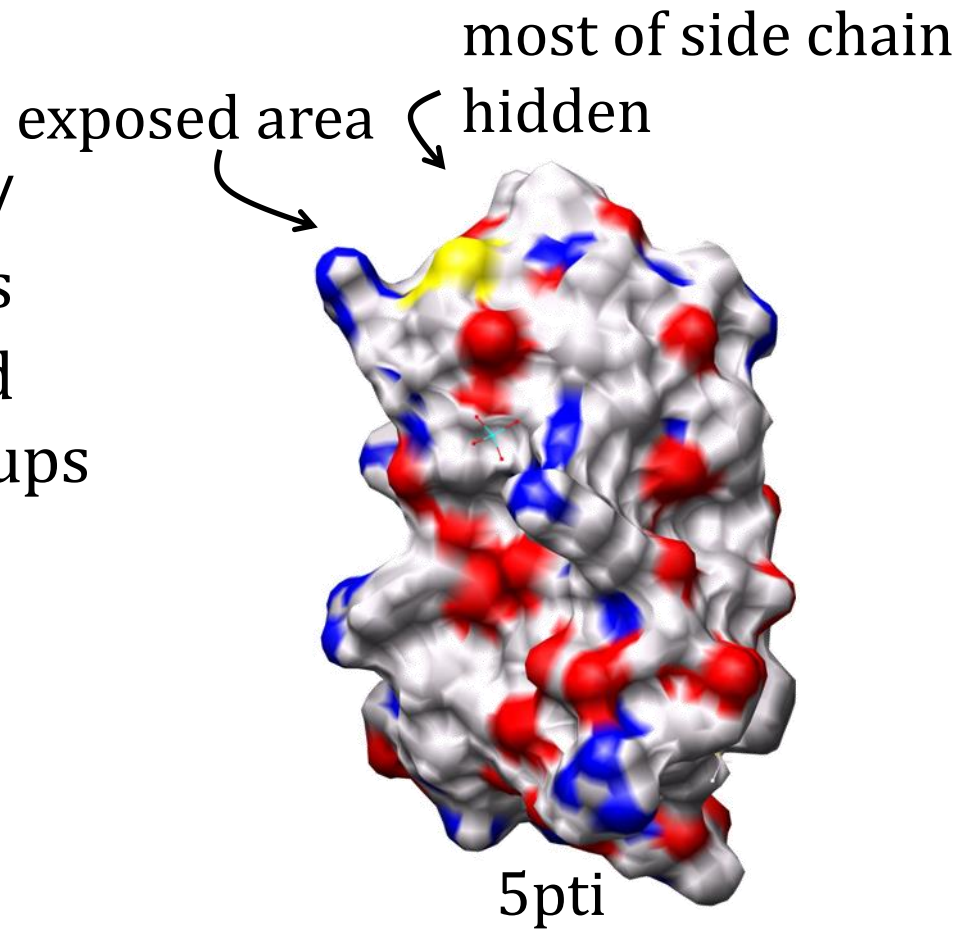
# Environment description

For each site measure the  $\text{\AA}^2$   
exposed to solvent

Sometimes one has charges /  
polar groups touching others

- measure fraction of buried  
area covered by polar groups

Define environments...

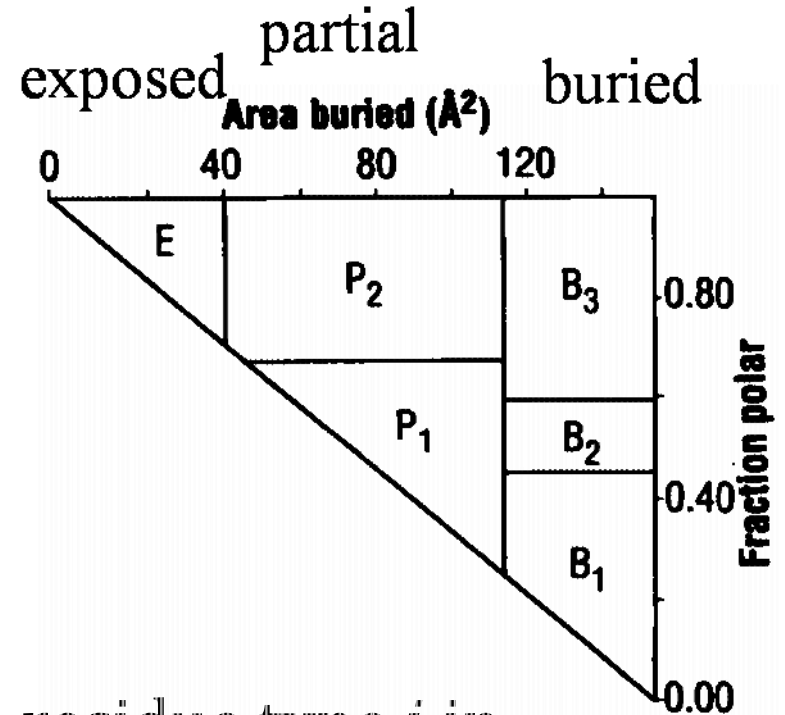


# Environment description

- 6 environment types
- 3 secondary structure types
  - $\alpha$ ,  $\beta$ , others
- = 18 environments

## Data collection

- 16 proteins
- find environment of each site
- count
  - how many times does one see residue type  $i$  in environment  $j = N(i, j)$
- count – how often does one see residue type  $i = N(i)$



# Environment description

How unusual is a residue  $i$  in environment  $j$ ?

$$score(i, j) = \ln \left( \frac{N(i, j)}{N(i)} \right)$$

Final result ? a big scoring table

likely

unlikely

what one expects

Environment class	W	F	Y	L	I	V	M	A	G	P	C	T	S	Q	N	E	D	H	K	R
B <sub>1</sub> α	1.00	1.32	0.18	1.27	1.17	0.66	1.26	-0.66	-2.53	-1.16	-0.73	-1.29	-2.73	-1.08	-1.93	-1.74	-1.97	-0.34	-1.82	-1.67
B <sub>1</sub> β	1.17	0.85	0.07	1.13	1.47	1.09	0.55	-0.79	-2.02	-0.94	-0.22	-1.12	-2.91	-1.67	-1.42	-1.93	-2.56	-1.91	-2.69	-1.16
B <sub>1</sub>	1.05	1.45	0.17	1.10	1.11	1.02	0.98	-0.91	-1.92	0.26	-1.22	-1.53	-2.81	-1.17	-2.42	-2.52	-1.76	-1.12	-2.59	-2.16
B <sub>2</sub> α	0.50	0.90	0.85	1.01	0.83	0.68	1.12	-0.89	-1.49	-2.21	-0.10	-1.50	-1.47	-0.23	-0.81	-0.71	-1.62	0.23	-0.78	0.06
B <sub>2</sub> β	0.01	1.18	1.06	0.76	1.31	1.06	0.64	-1.55	-2.26	-0.49	-0.87	-2.27	-1.77	-1.22	-2.07	-1.07	-1.41	-0.77	-1.14	-0.20
B <sub>2</sub>	1.02	1.05	1.12	0.84	0.81	0.60	0.90	-0.66	-1.66	0.19	-0.05	-0.76	-1.17	-0.76	-0.66	-1.35	-1.28	0.46	-2.34	-0.80
B <sub>3</sub> α	0.92	-0.03	0.58	0.15	0.04	-0.02	0.89	-0.57	-1.86	-0.68	-1.56	-0.57	-0.96	0.22	-0.06	0.08	-0.50	0.73	0.43	0.96
B <sub>3</sub> β	0.75	0.81	1.30	0.18	0.54	0.56	-0.57	-0.93	-1.93	-0.34	-0.54	-0.44	-0.74	0.21	-0.24	-0.14	-0.86	0.82	-0.53	0.13
B <sub>3</sub>	1.07	0.70	1.13	0.35	-0.17	-0.03	0.23	-0.96	-0.98	-0.13	-1.20	-0.53	-0.54	0.05	0.04	-0.36	-1.05	1.01	0.10	0.66
P <sub>1</sub> α	-1.35	-0.82	-0.59	-0.52	-0.24	0.10	-0.03	0.73	-0.49	-0.25	0.95	0.31	0.34	-0.14	-0.54	-0.17	-0.25	-0.52	-0.21	-0.28
P <sub>1</sub> β	0.36	-0.49	0.17	-1.03	0.20	0.46	-0.27	0.64	-0.82	-0.55	1.49	0.93	0.33	-2.27	-1.32	-0.73	-1.07	-0.42	-1.21	-0.77
P <sub>1</sub>	-1.26	-1.20	-1.31	-0.62	-0.23	-0.01	-1.19	0.46	-0.24	0.66	1.35	0.56	0.49	-0.63	-0.13	-0.61	0.38	-1.12	-0.74	-1.29
P <sub>2</sub> α	-1.14	-1.43	-0.79	-0.35	-0.54	-0.48	-0.45	0.06	-0.50	-0.26	-0.93	-0.05	-0.18	0.55	-0.05	0.56	0.28	0.06	0.61	0.50
P <sub>2</sub> β	-0.79	-0.54	-0.84	-1.30	-0.33	0.13	-0.72	-0.55	-0.98	-1.29	0.84	0.59	-0.08	-0.16	0.32	0.19	-0.87	0.59	0.10	0.10
P <sub>2</sub>	-0.82	-0.86	-0.51	-0.70	-1.09	-0.88	-0.89	-0.15	-0.40	0.44	-0.60	0.06	0.26	0.27	0.50	0.27	0.49	0.13	0.44	0.30
E α	-1.35	-2.20	-2.10	-1.58	-2.76	-1.10	-0.72	0.46	0.68	0.04	-0.44	-0.17	0.15	0.36	0.28	0.59	0.44	-0.19	0.13	-0.34
E β	0.64	-0.90	0.30	-1.66	-1.47	-1.74	-0.68	0.06	1.46	-0.96	-0.24	0.14	0.65	-0.19	-0.06	-0.16	-0.78	-0.83	-0.52	-0.49
E	-2.14	-1.90	-0.94	-1.19	-1.61	-0.91	-1.67	0.12	1.13	0.20	-0.46	0.12	0.32	-0.03	0.41	0.03	0.22	-0.25	-0.14	-0.32



# Environment description - application

- given these descriptions – use them
- take a protein structure label each site
- take sequence of interest
- for each residue
  - score at each site of protein
- score matrix
- find best path
  - sequence to structure alignment

## Final application

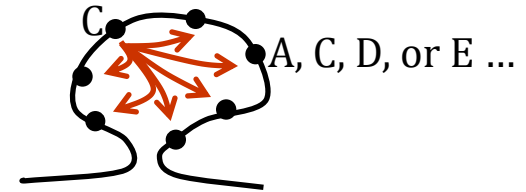
- take protein databank
- try to align your sequence to every structure

		sequence →						
		A	C	D	E	F	G	
structure sites	1		?					
	2							
	3							
	4							
	5							
	6							
	7							

# Frozen approximation

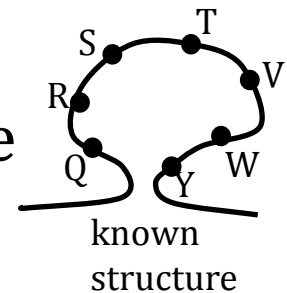
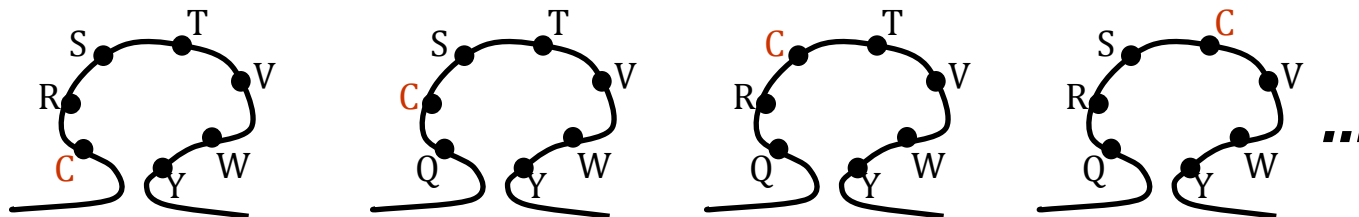
## Original problem

- we want to use a score function which
  - sensitive to sequence
  - sensitive to structure



Remember – original structure did have a sequence

- belief
  - if two proteins are related, the sequences will have similar properties
  - score with the residues of the original sequence



# Frozen approximation

I can score my sequence in the environment of the known structure

- good
  - the environment is well characterised
    - if my structure has polar residues here, they will go into the scoring function
- bad ?
  - we use the sequence of template (known structure)
  - it may only allow very related residues
  - original aim was to move away from close sequences

		sequence						
		A	C	D	E	F	G	
structure sites	1	?						
	2							
	3							
	4							
	5							
	6							
	7							

# Summary so far

- look for closely related templates
- try sequence based methods
- sequence to structure methods are definitely possible
- can I make better scoring schemes ?

# Scoring schemes

... S T D G W Y F I L S T ...  
polar / charged | small | hydrophobic | polar

- how much structural information is hidden in sequence ?
- look at a sequence
- I already have labels for sites
  - implicit in substitution matrices
- does the structure contain extra information ? ...

# Extra information from structures

Residues exist in a protein for different reasons

- gly is easy to substitute – look at diagonal in blosum matrix
- in some turns, gly is essential
  - can only be seen from structure
- cys
  - sometimes a normal hydrophobic residue
  - sometimes the geometry says it must form a disulfide bond
  - structure can say if there is another cys near in space
- ...
- it should be useful to combine sequence and structure information

# Extra information from structures

Claim – hope

- combination of methods has better signal / noise

Implementation ? easy in principle

for each residue  $i$  in your query sequence

for each site  $j$  in template

calculate sequence score  $s_1$  based on profile of  $i$

calculate structural score  $s_2$  based on fitting residue type  $i$  into site  $j$

score for alignment matrix =  $s_1 + k s_2$

for some constant  $k$

# In practice

- most fold recognition programs combine sequence terms and structural scores
- results may or may not be better than best pure sequence methods
- problems..



# Problems with clever methods

## Simple sequence searches

- good models for statistical significance
  - (is a related protein really related ?)

## Remote sequence searches (psi-blast)

- statistics OK, but less reliable

## Structure / Sequence+structure methods ?

- no good model for scores
- no good model for statistical significance

## how will score grow with

- size of query ?
- size of alignment ?
- sequence composition ?

# Principle

If you have extra information (structure)

- must be a good idea to use it

	sequence	structure based
database size	$4 \times 10^7$	$10^5$
	fast	slow
scores	good models	weaker
statistical significance	good or almost good	weaker

# Summarise and stop

- Use sequence information when possible
- use adventurous sequence methods when necessary
- use very speculative methods (sequence to structure) when necessary