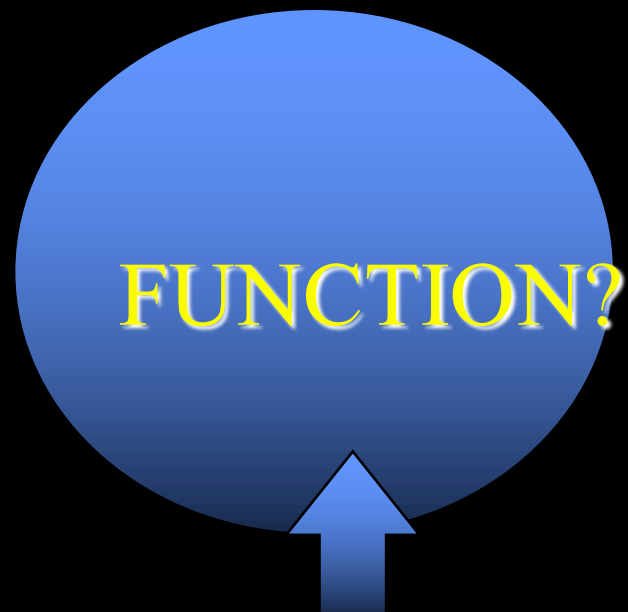
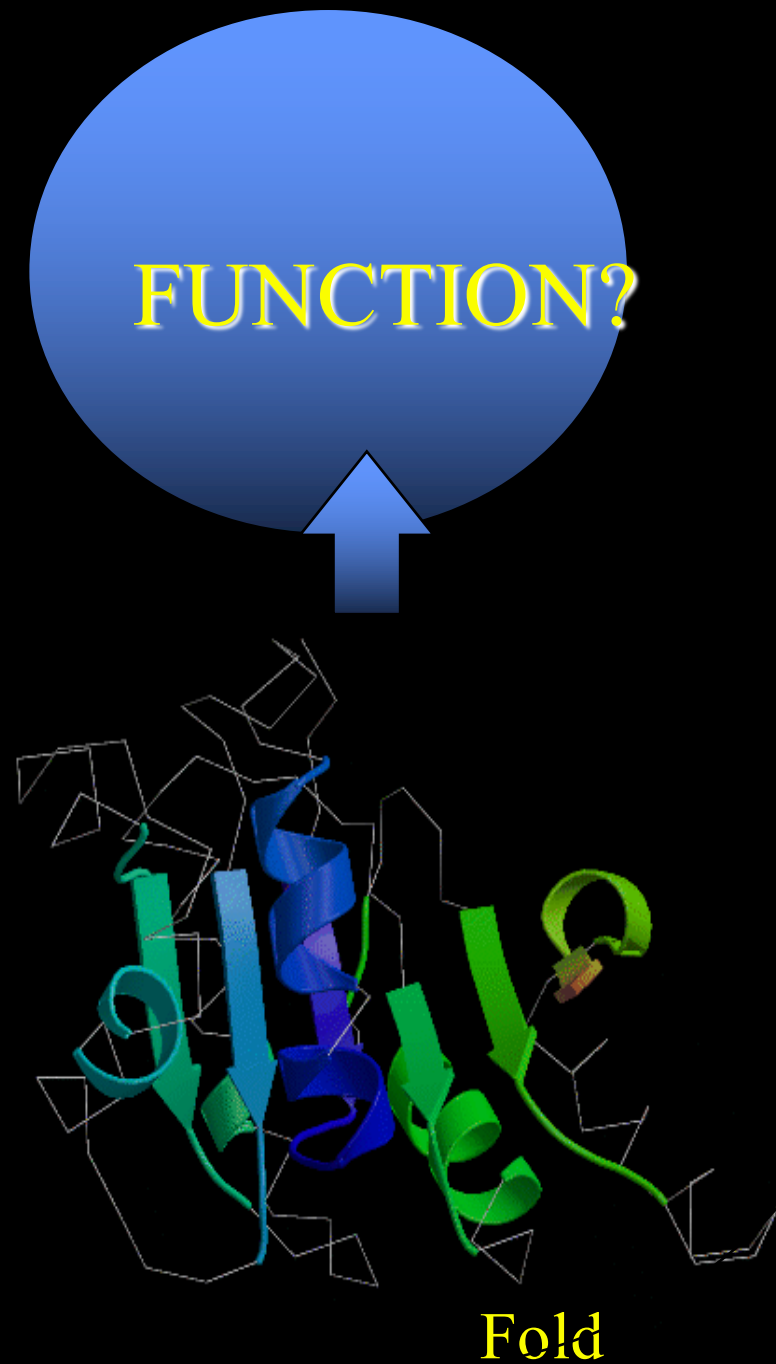
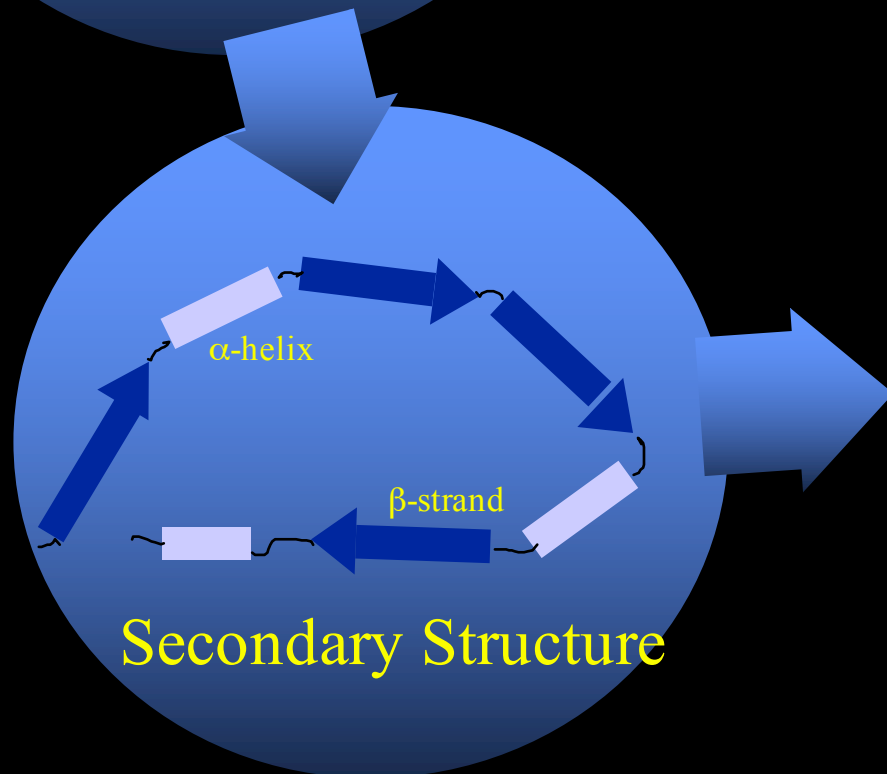
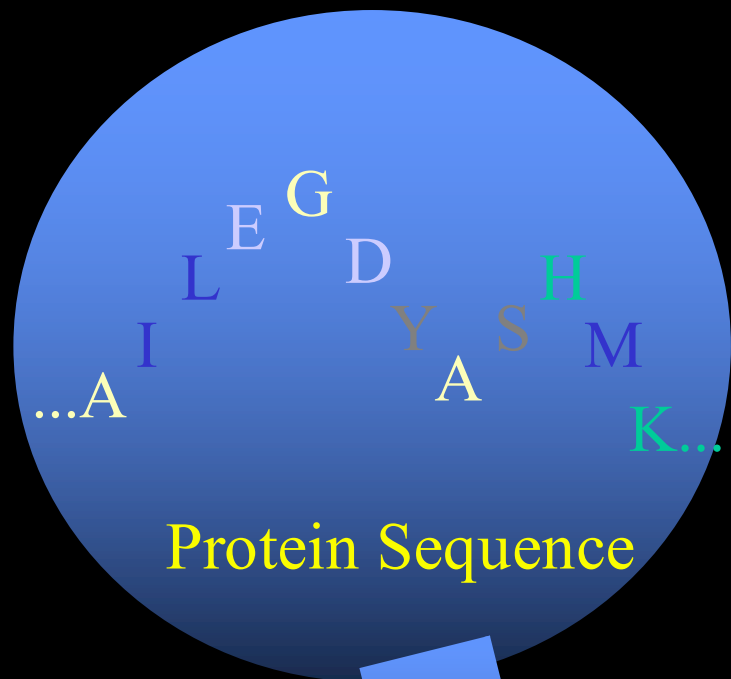


# **Protein Structure Prediction**

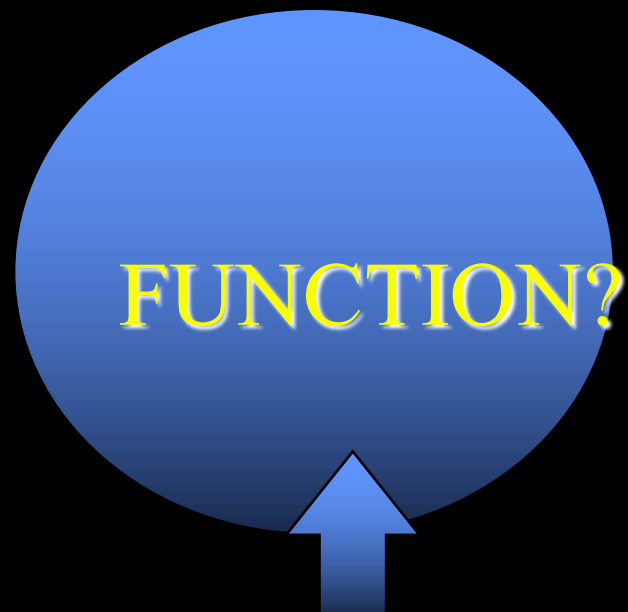
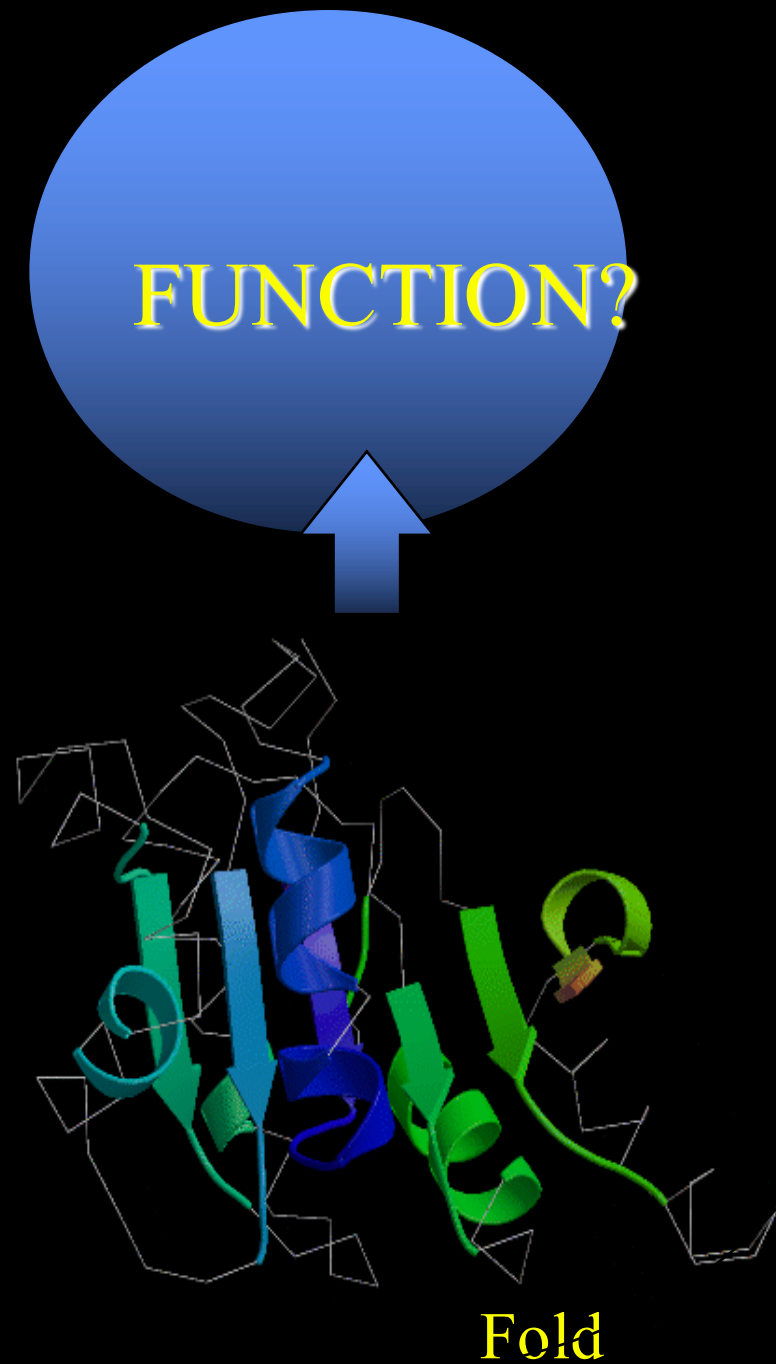
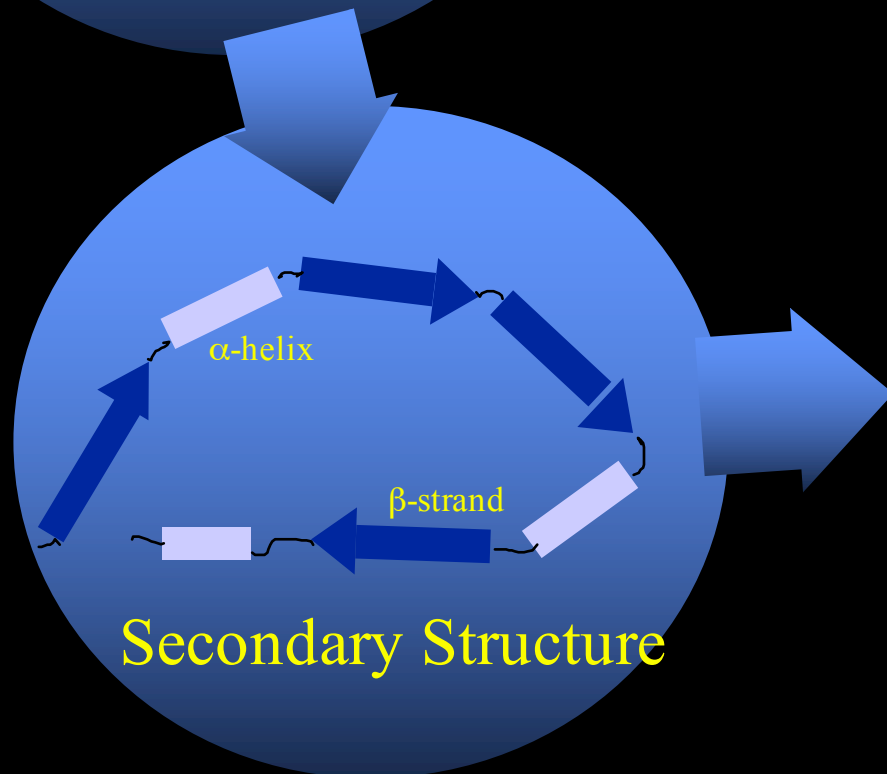
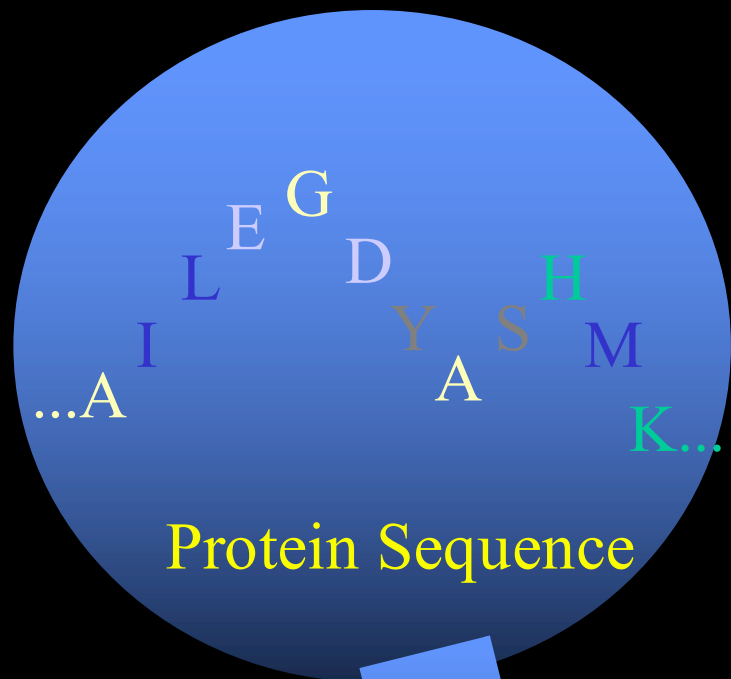
Geoff Barton

[www.compbio.dundee.ac.uk](http://www.compbio.dundee.ac.uk)



# Protein Structure Prediction

- Since all the information about the protein three-dimensional structure is contained in the protein sequence, it should be possible to predict the protein three-dimensional structure given only the sequence...
- However, this turns out put be pretty hard to do...





**Before talking about  
prediction...**

# Protein Structure Determination by Experiment

- From X-ray crystallography
- From NMR spectroscopy
- From Electron Microscopy

# Protein Structure Determination

- All structures available are 'models'.
- A model is simply a hypothesis based on the available data.
- THE MORE DATA YOU HAVE, THE MORE RELIABLE THE MODEL

# X-ray crystallography

For *small molecules* (not macromolecules)

- At 0.8 Angstrom resolution
  - 10 observations/parameter (9 parameters)
  - i.e. **90 'measurements'** per atom.
- At 1.2 Angstrom resolution
  - 5 observations/parameter (4 parameters)
  - i.e. **20 'measurements'** per atom.

# X-ray crystallography

## For proteins ...

- At 2.0 Angstroms resolution
  - 2 observations/parameter (4 parameters)
  - i.e. **8 measurements per atom**

Number of observations is improved by including **geometry restraints** from knowledge of protein structure.

# NMR Spectroscopy

- Good small molecule X-ray structure has 90 measurements per atom.
- Reasonable protein X-ray structure has 20 measurements per atom.
- **Good NMR protein structure has 10 measurements *per residue*.**

# NMR Spectroscopy *versus* X-ray

- NMR has fewer measurements per residue than X-ray crystallography on which to base the model.
- NMR techniques rely more heavily on basic geometry of amino acids and protein backbone.

**So, what about structure prediction?**

# Methods for protein structure determination

Method	Measurements per residue
X-Ray crystallography at 2.0 Angstroms	80 + geometry
Good NMR structure	10 + geometry
<i>Ab-initio</i> prediction	0 + geometry!!



# Protein structure prediction

Since there are 0-few measurements, the challenge is to find constraints on the structure from somewhere...

- Homology to known structure.
- Distance restraints (disulphides etc).
- Restraints from secondary structure prediction.
- Buried/Exposed restraints.

# ***Ab initio* structure prediction**

# *Ab initio* prediction - energy calculations

- Attempt to 'fold' protein from extended chain.
- Model all inter-atomic interactions as an energy term.
- Find the conformation that minimises the sum of pair-wise interaction energies.

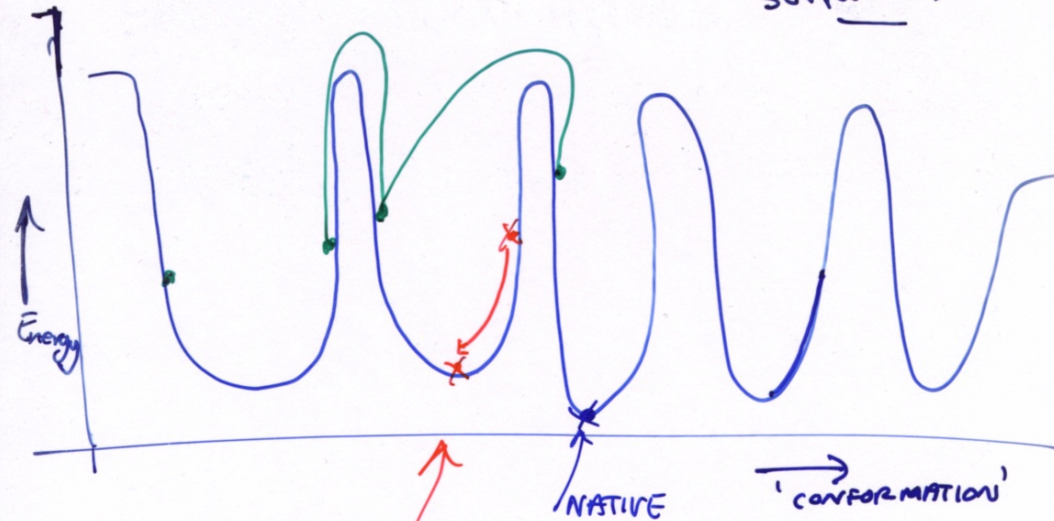
①

## ENERGY CALCULATIONS



UNFORTUNATELY

— EGG BOX  
Potential energy  
surface.



→ ENERGY MINIMISATION

MOLECULAR DYNAMICS

SA. - LARGE RADIUS OF CONVERGENCE

# *Ab initio* - energy calculations

## Problems

- Potentials.
- Search space is large.
- Multiple minima - distinguishing native from non-native conformations

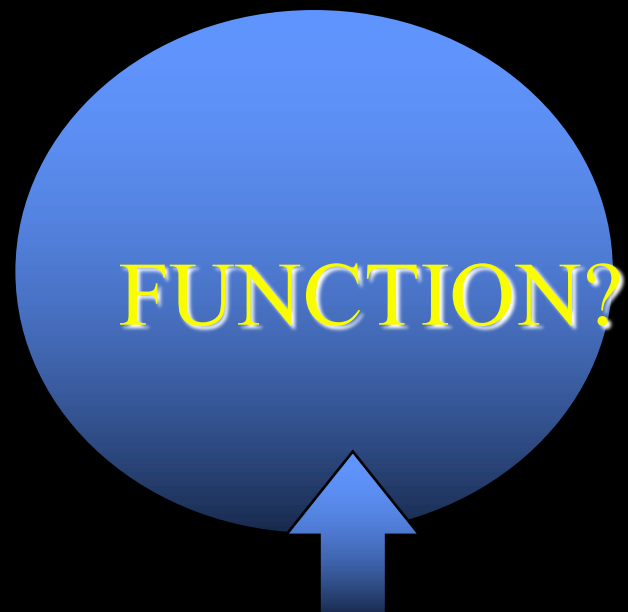
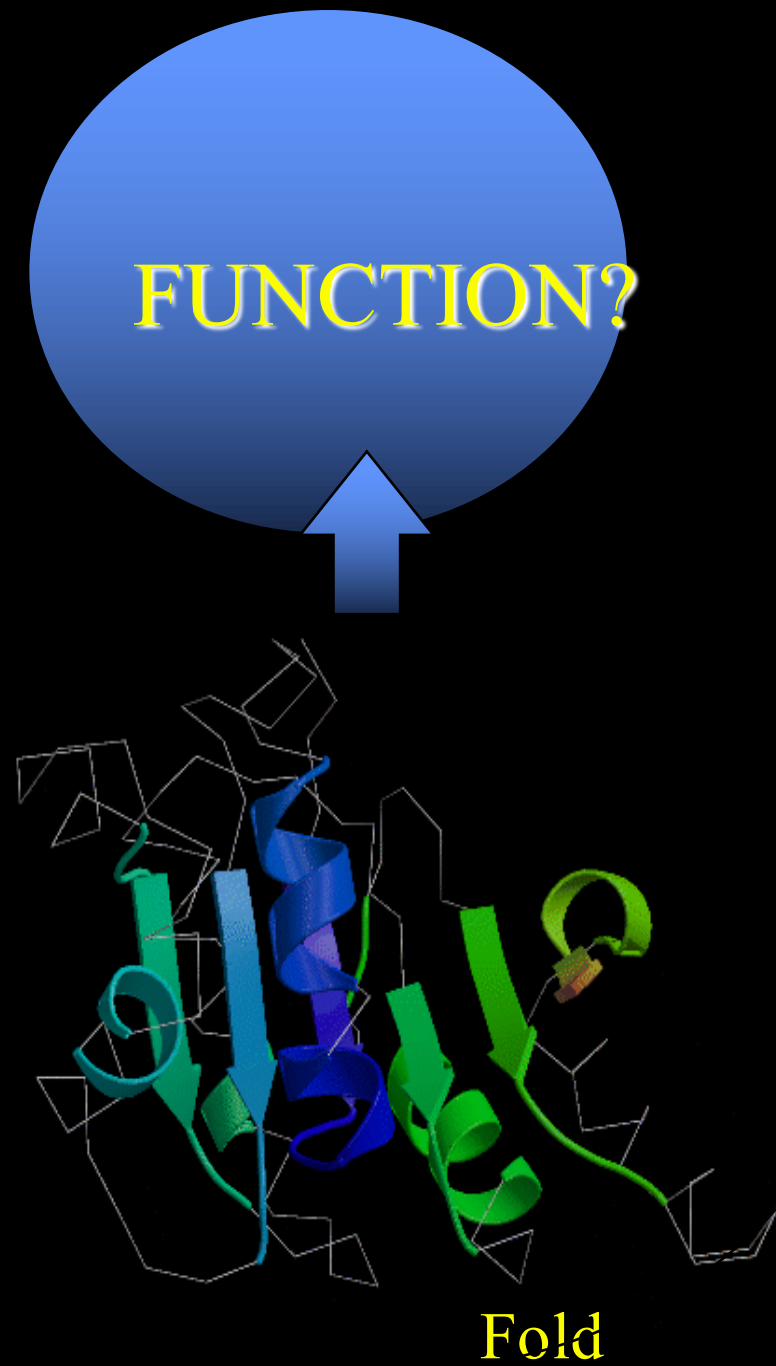
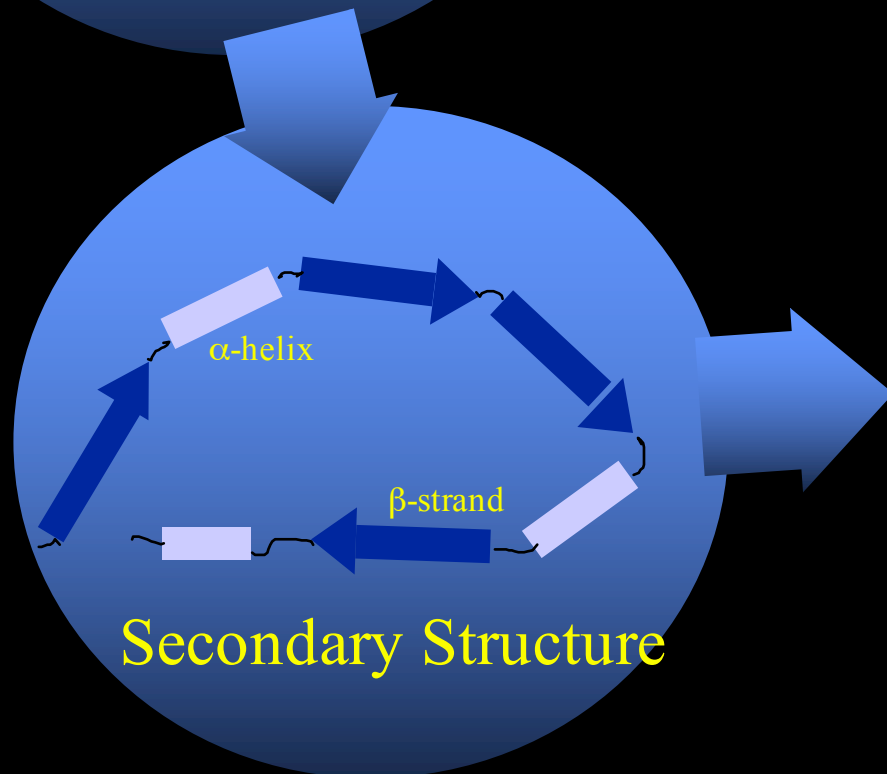
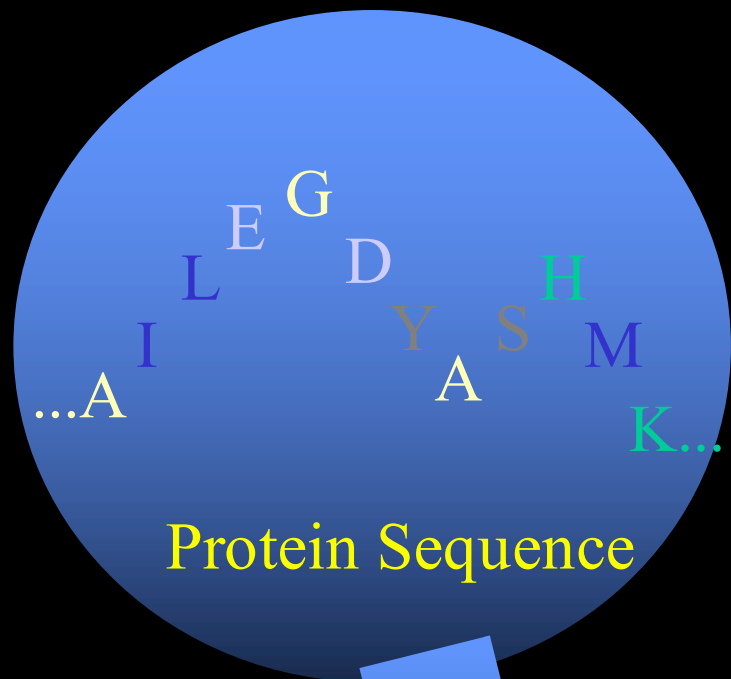
# *Ab initio* - energy calculations

- So far only limited success in conjunction with strong restraints. For example:
  - Coiled coils.
  - Disulphide rich proteins.

# Hybrid *Ab initio* methods

- Fragment assembly – David Baker – Rosetta and “folding at home” – see Baker Lab web site.
- Generate large number of small fragments of structure consistent with a simplified potential and known folding units
- Cluster millions of generated structures to find most consistent folds.
- Also see *Nature* 8<sup>th</sup> Nov 2007 for application in X-ray crystallography.





# Protein Structure Prediction

## *Homology modelling*

- **Exploits principle that “similarity of sequence implies similarity of 3D structure”**
- Relies on finding similarity to protein of known structure.
- Must get alignment between known and unknown correct.
- Build loops correctly.

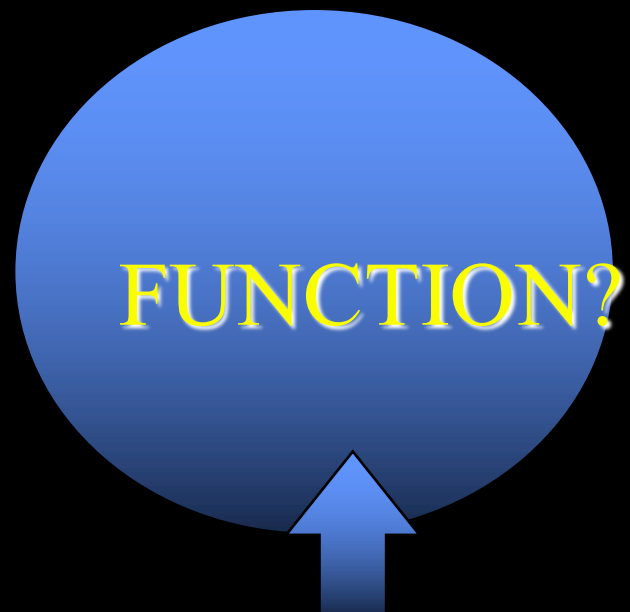
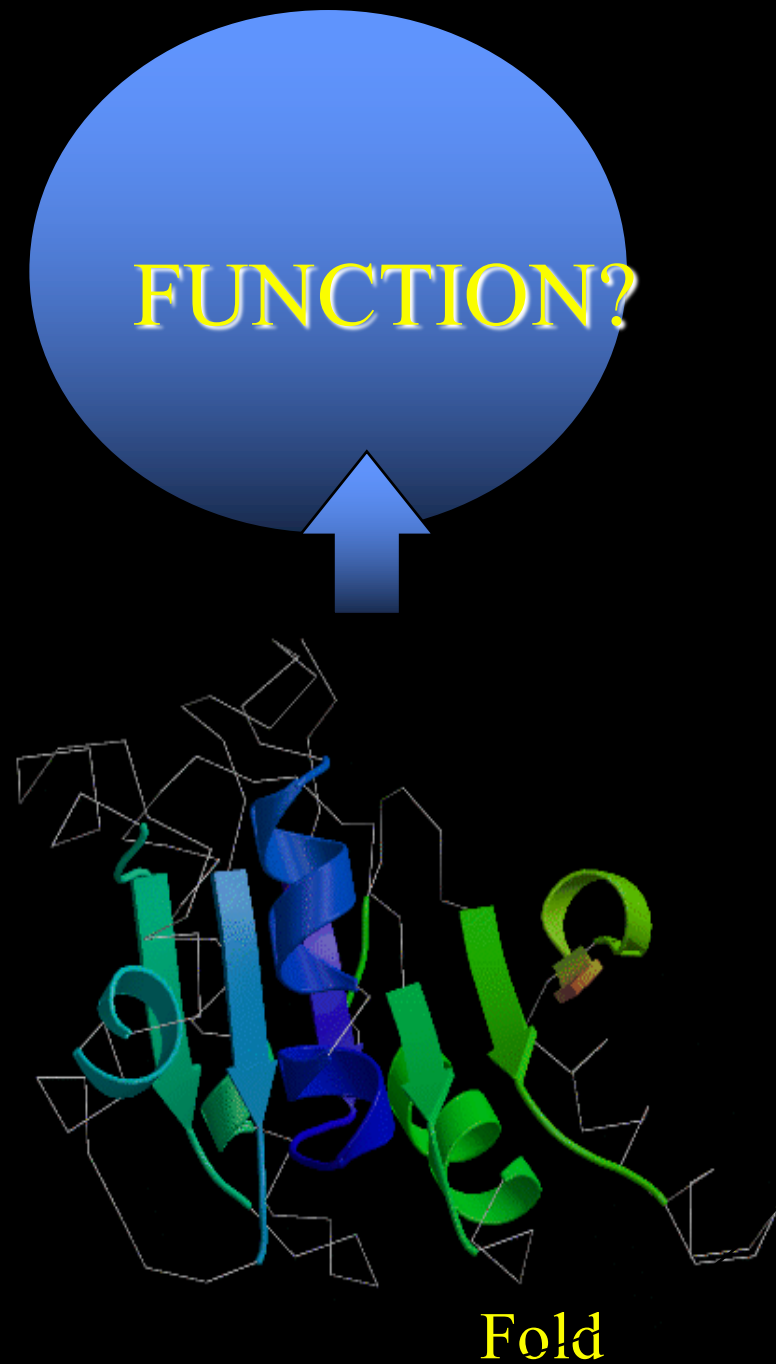
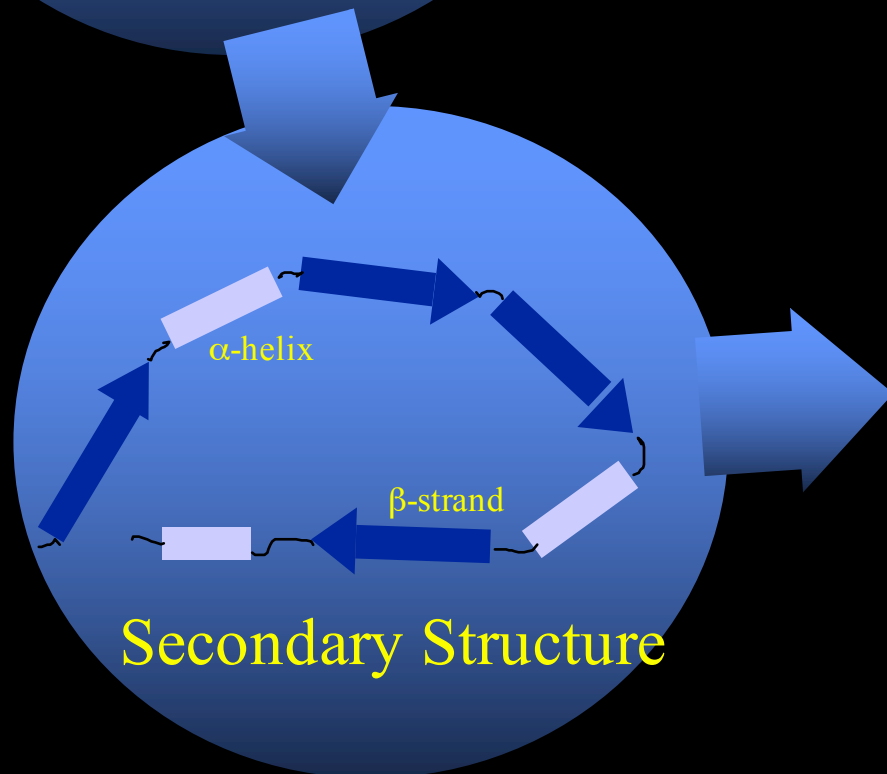
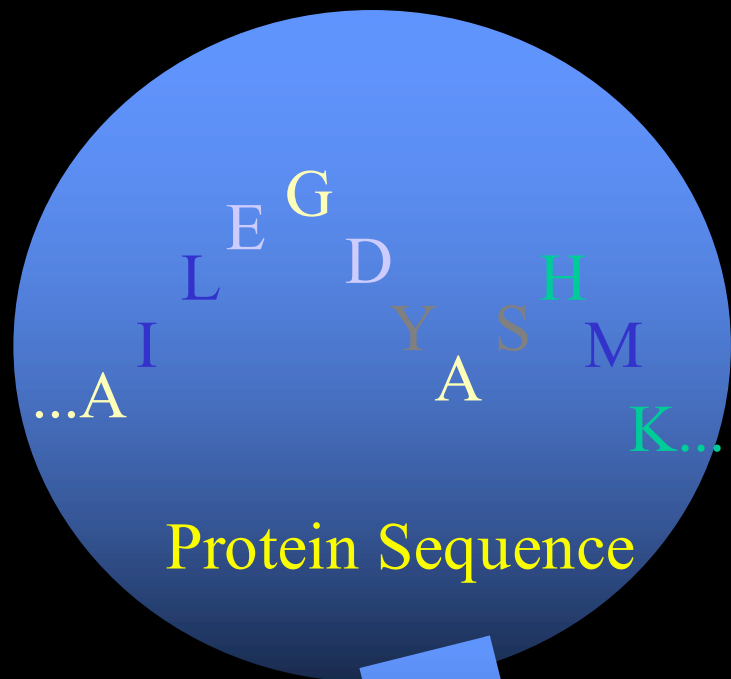
# Resources for Homology Modelling

- Swissmodel
  - <http://swissmodel.expasy.org/>
- ModBase
  - <http://modbase.compbio.ucsf.edu>
- PhyreII
  - Imperial College – Google it!

# Homology Modelling

1. Find homologue of known structure (scaffold)
2. Accurately align target sequence to scaffold protein sequence
3. Substitute amino acids in conserved core of scaffold protein for those in the target
4. Adjust side-chain conformations to remove clashes and optimise geometry
5. Model loops in the target that are of different length to those in the scaffold
6. Build multiple models where there is uncertainty over any of the above stages
7. Check model for gross errors (buried charges/exposed hydrophobics etc)

# **Protein Secondary Structure Prediction**



# Protein Secondary Structure Prediction

Useful, even if not used to gain a full three-dimensional structure.

Identify 'safe' regions for mutagenesis

Help improve sequence searches/fold recognition.

Improve sequence alignment accuracy.

# A Few Ideas That Have Been Tried in Protein Secondary Structure Prediction

- Sequence Similarity: Do a sequence alignment against a protein of known structure.
- By eye: Look for patterns of conservation in a multiple sequence alignment.
- Statistical Methods: Singlets (Chou & Fasman), Information theory:(GORI,II,II,IV...) Bayes etc. etc. etc....
- Pattern Matching: Templates (Taylor), AI (Cohen) etc. ...
- Neural Nets: Karplus, Rost & Sander, etc...
- Nearest Neighbour: Salamov & Solovyev, Levin, etc ...
- Others: Tarot, Tea Leaves etc ...



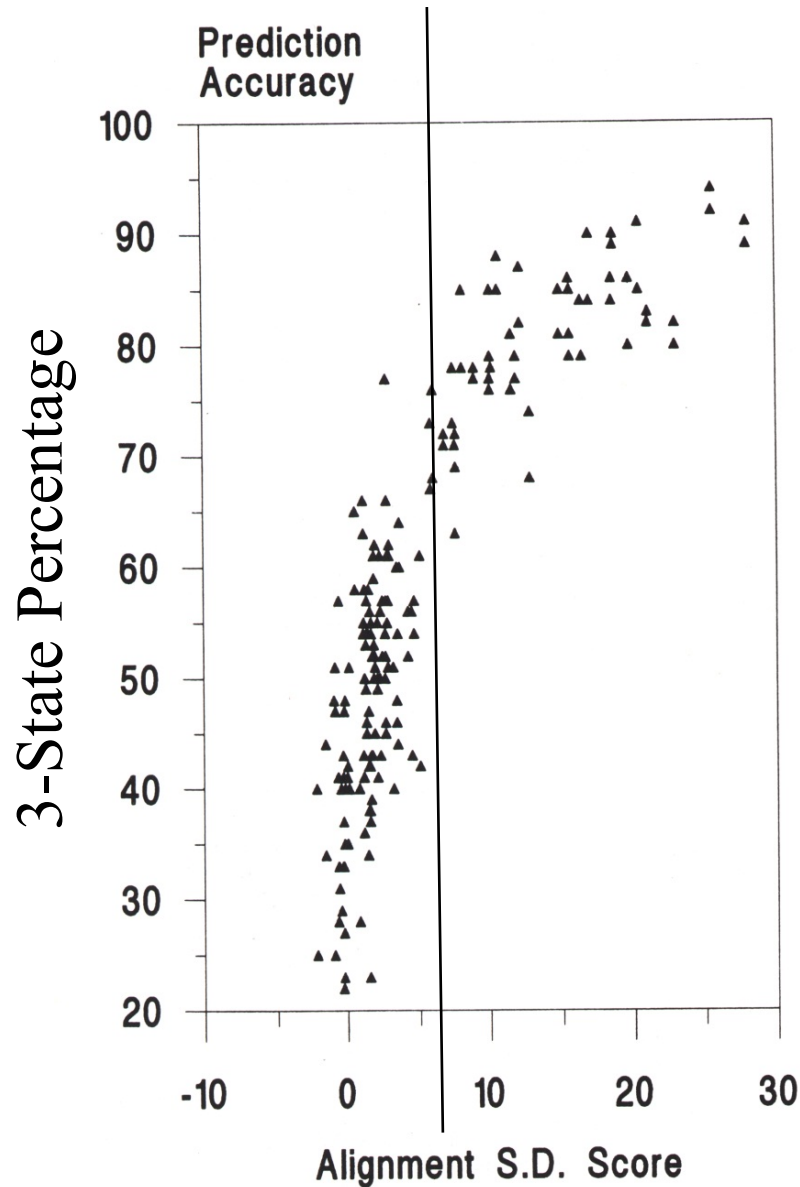
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**If the protein is similar to  
a protein of known three  
dimensional structure,  
then just use alignment...**



**Fig. 2.** The accuracy of secondary structure prediction by sequence alignment plotted against the alignment SD score to the homologous protein. One hundred and eighty-two predictions were made from pairwise alignment of the proteins in Table I.

**Prediction by alignment to protein of known structure.**

No poor predictions by alignment for similarity > 6 sigma.

Boscott, P. E., Barton, G. J. and Richards, W. G. (1993), *Prot. Eng.*, **6**, 261-266.

# Prediction methods normally need

- To be trained on some data
- Tested on different data

# For secondary structure prediction:

- Train on a subset of the Protein Data Bank (PDB) of known protein three-dimensional structures
- Test on different subset of the PDB that was not used in training
- Best test is on proteins for which no structure is known.
  - However, you have to wait for someone to solve the structure by X-ray methods before you can find out the result

# A Few Ideas That Have Been Tried in Protein Secondary Structure Prediction

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# Statistical Methods

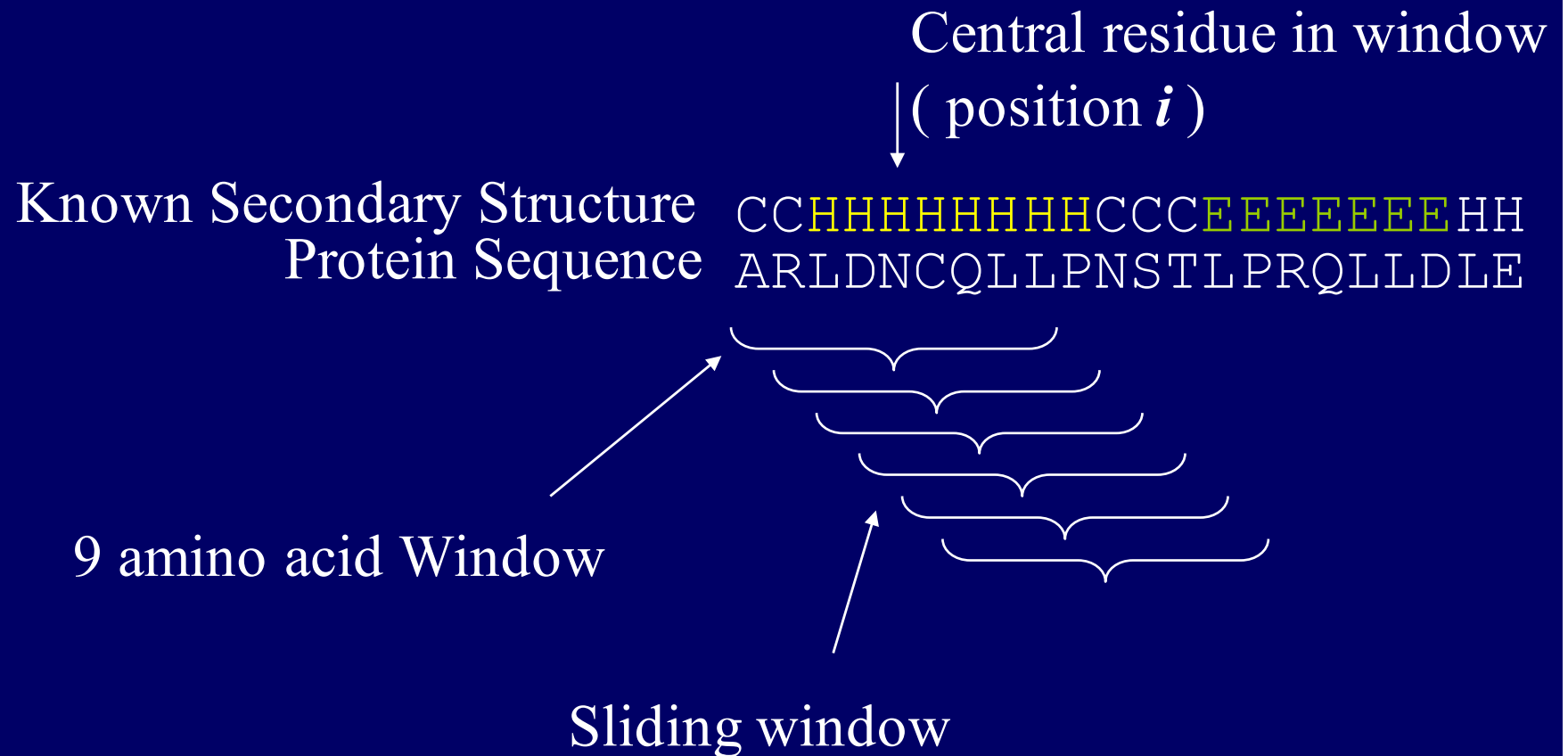
- Simplest method - singlets:
  - Look at proteins of known 3D structure
  - Count how many times each amino acid is seen in each secondary structure state. For example, Ala in Helix, Strand, Turn etc
  - Express as a probability that each amino acid will appear in that state



# Statistical Methods

- More complex method - windowing:
  - Take a sliding window of say, 9 amino acids
  - Gather statistics on all 9 positions relative to the secondary structure state of the central position
  - Express as a probability or “information content”

# More complex method - windowing



Gather statistics about how often each amino acid is seen at positions  $i-4, i-3, i-2, i-1, i, i+1, i+2, i+3, i+4$

**For each secondary structure type at position  $i$ .**

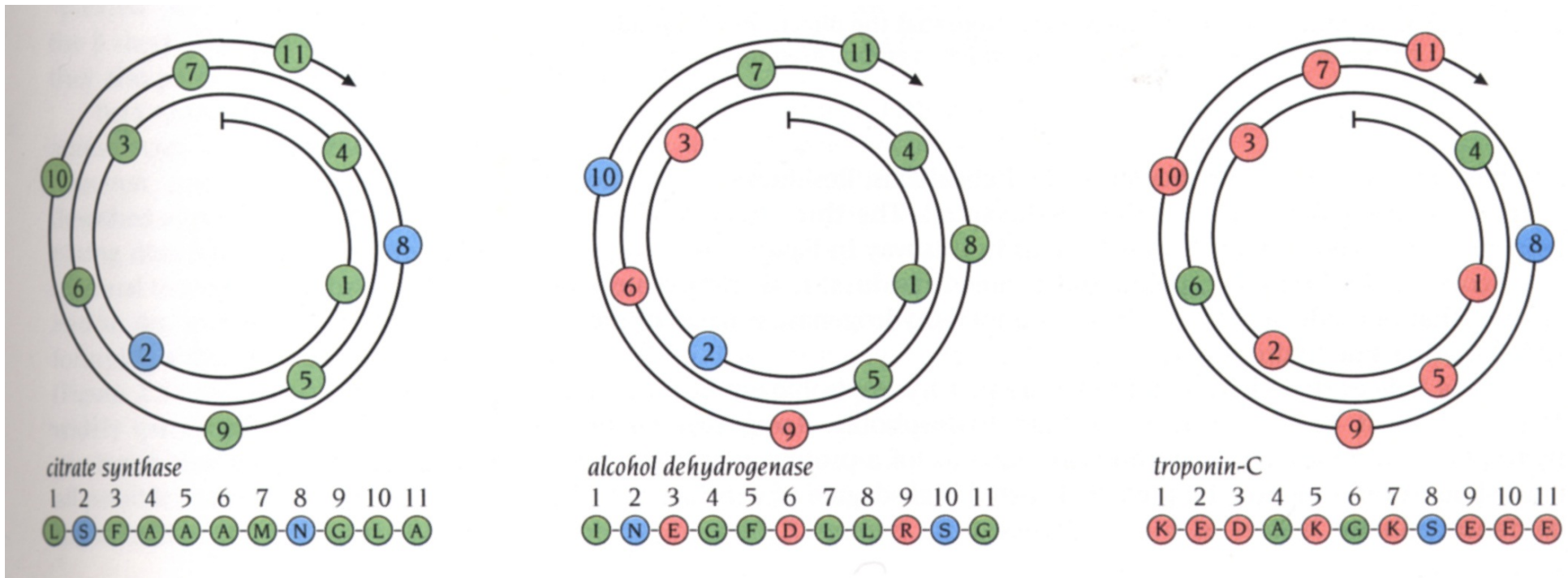
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1. Buried helix; 2. part exposed helix; 3. exposed helix

Helical wheel plots to show location of hydrophobic amino acids on face of helix.



# The **more data** you have, the better the prediction

- Secondary structure predictions from *multiple sequence alignments* are more accurate than from single sequences.

Some numbers later...

- For now, some example blind predictions.

# Why predict by eye when we have programs to do this?

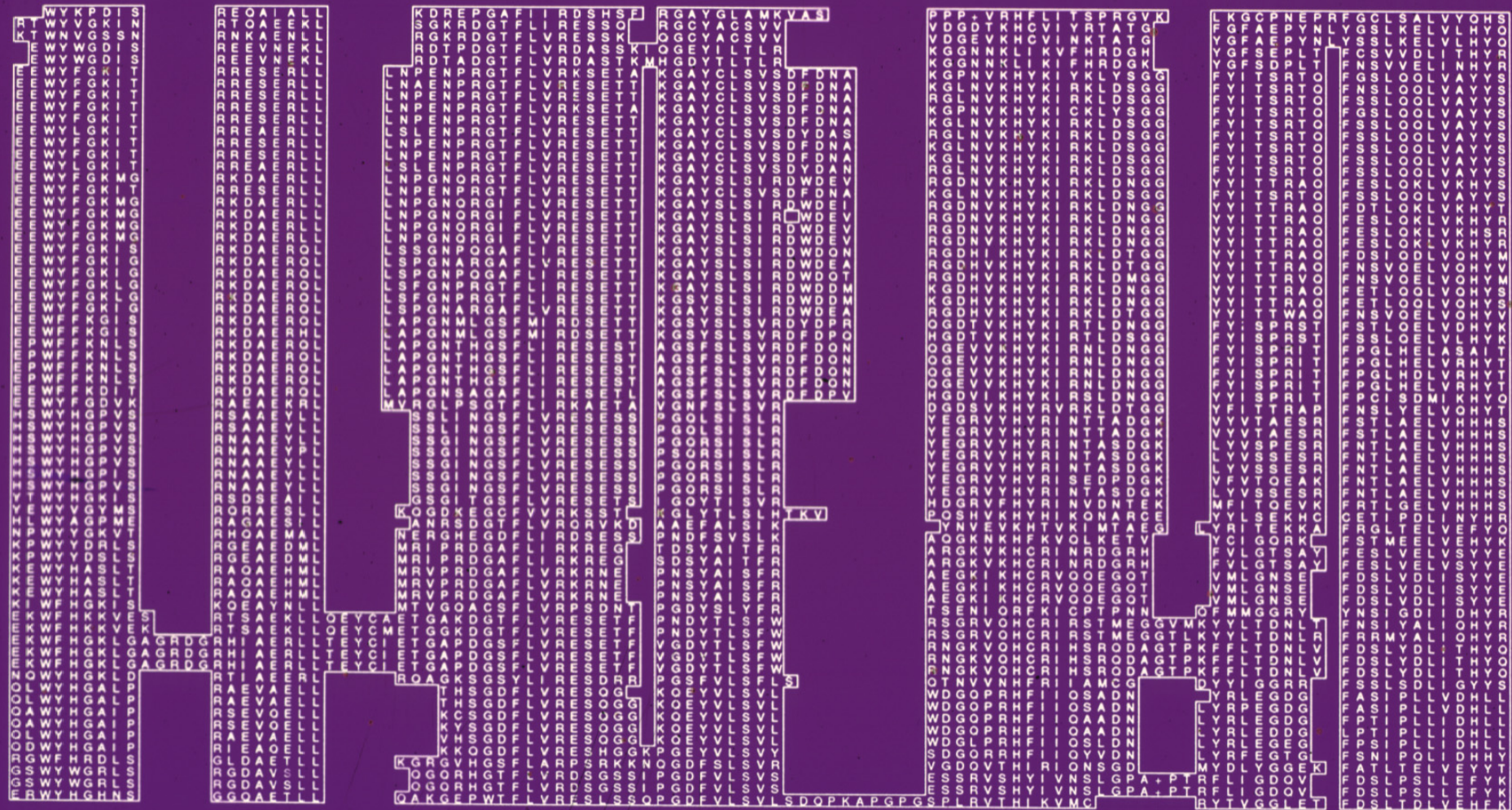
- Predicting by eye, helps to understand the evidence upon which an automatic prediction is based.
- Examples, from early work.

# Example Prediction

- SH2 Domains – phosphotyrosine binding domains seen in many signalling proteins



## SH2 alignment - 67 sequences



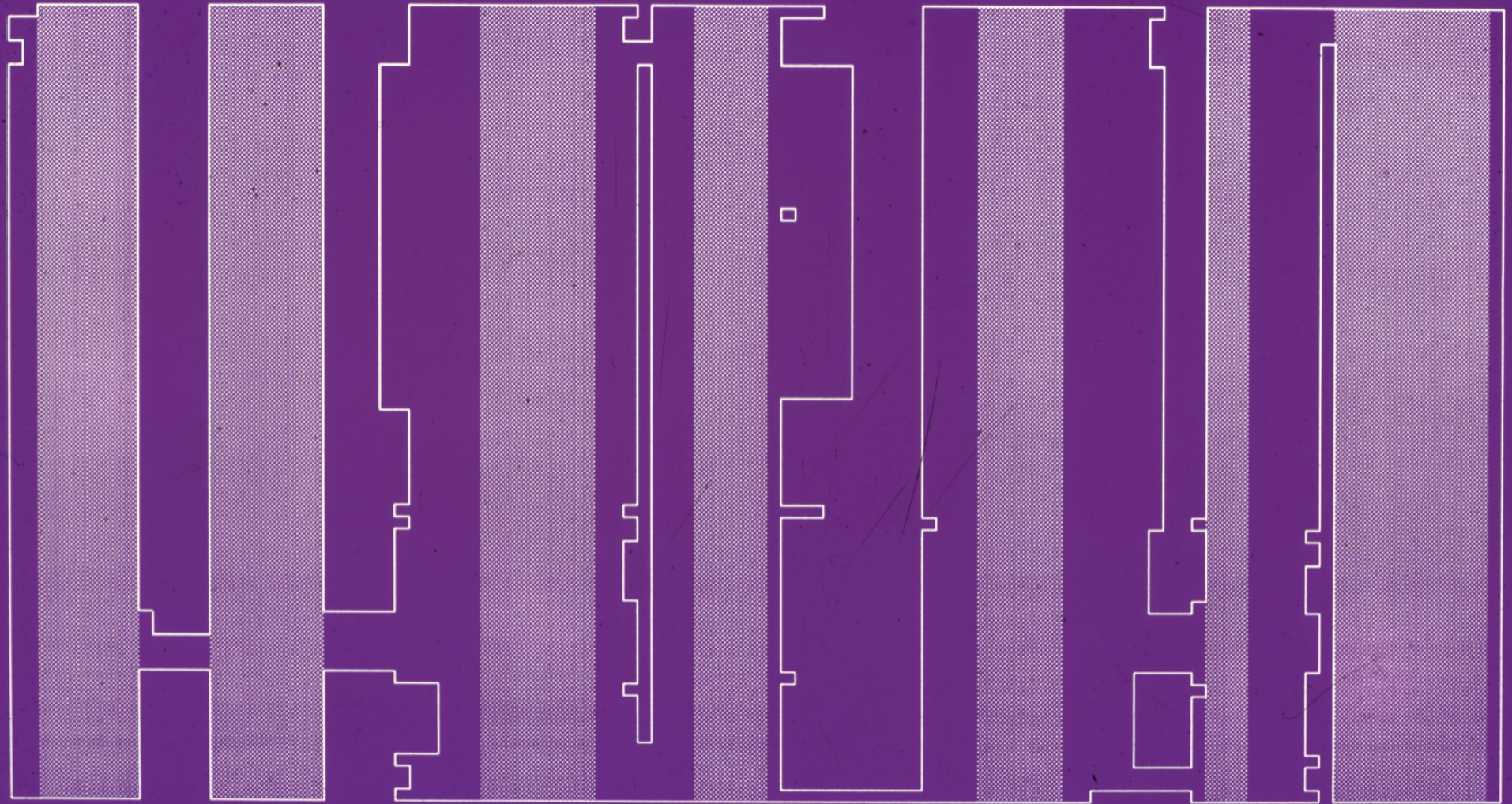
Russell, R. B. Breed, J. and Barton, G. J. (1992), *FEBS Lett.*, **304**, 15-20.



## SH2 alignment - without the sequences

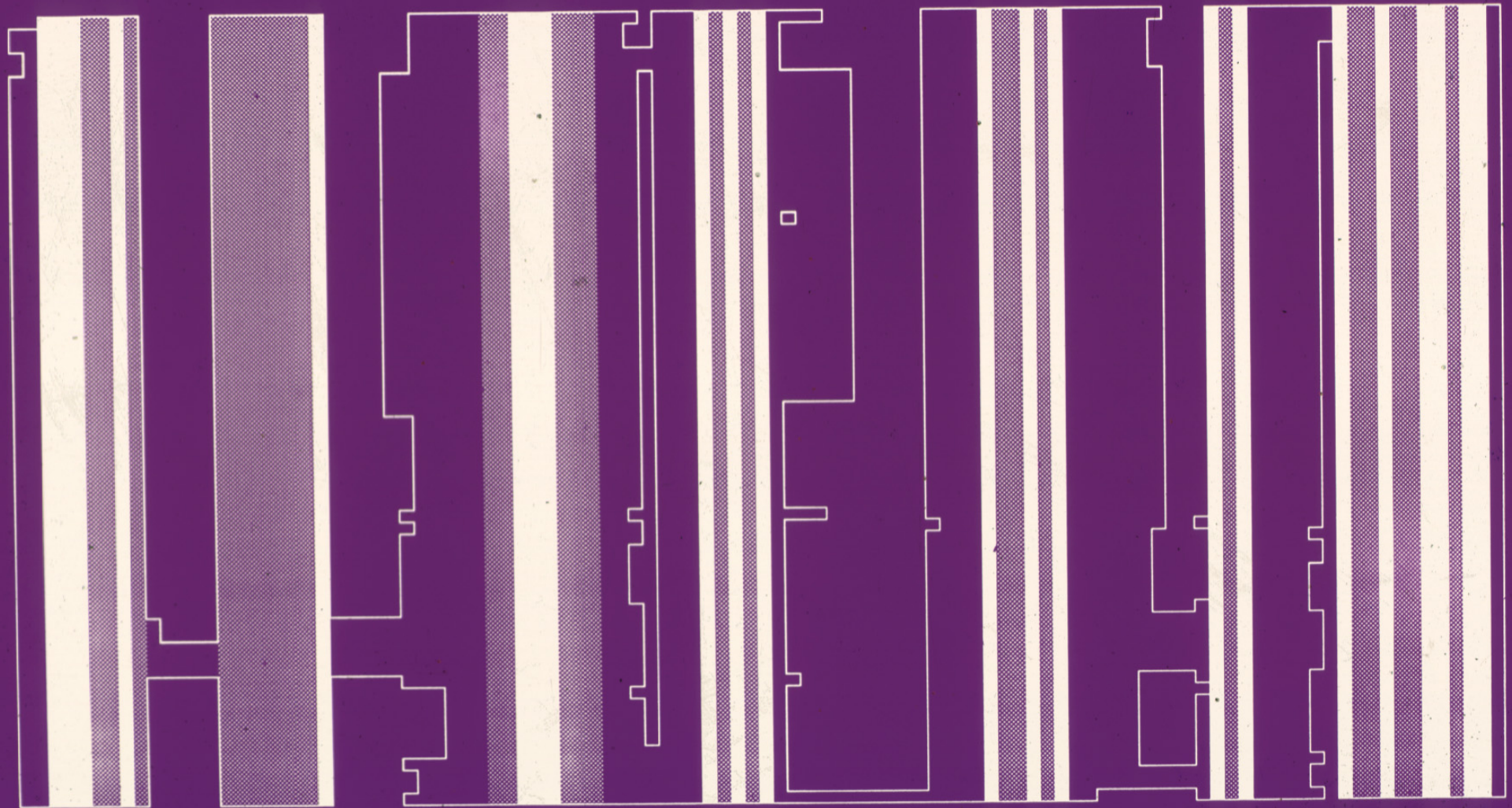


## SH2 alignment - conserved regions

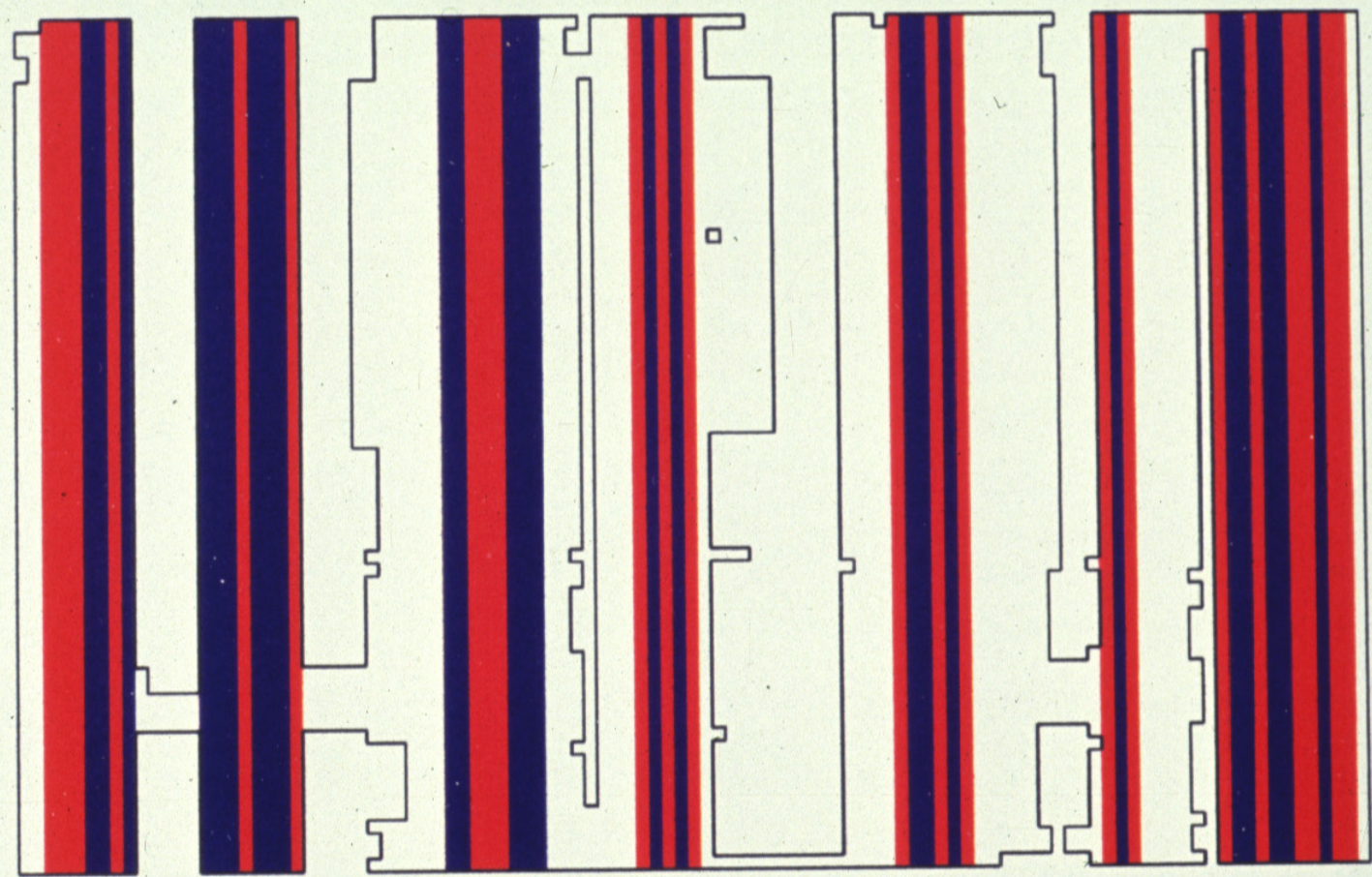




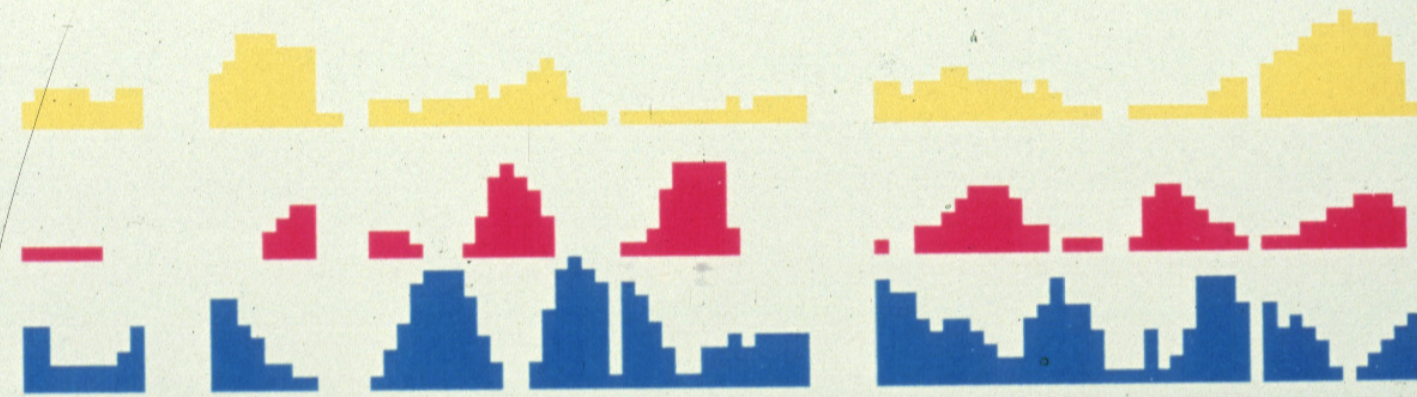
## SH2 alignment - conserved regions and hydrophobics







Conserved  
Hydrophobic



Helix  
Strand  
Turn





1, 4, 5, 8 conservation - surface helix

Short Strand possibly buried

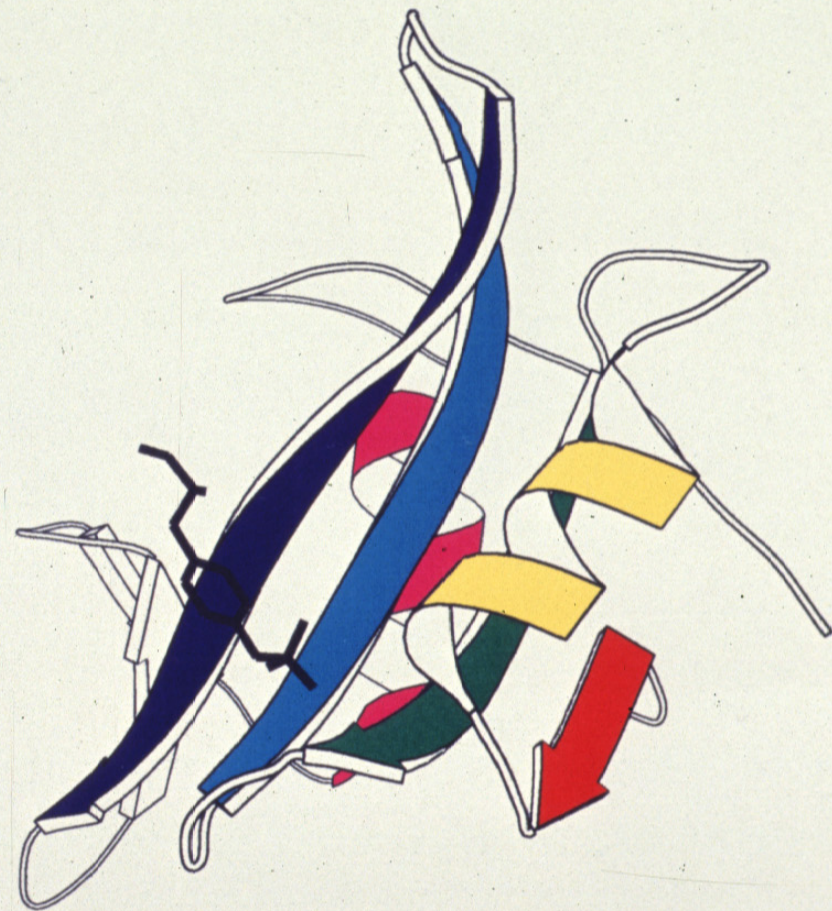
Alternating hydrophobics suggest surface strand

Alternating hydrophobics suggest surface strand

Buried strand with exposed face at end

No clear hydrophobic pattern

Run of 3 conserved hydrophobics suggest strand



## src SH2-Domain

(Waksman et al, Nature, 358, 646-653, 1992)

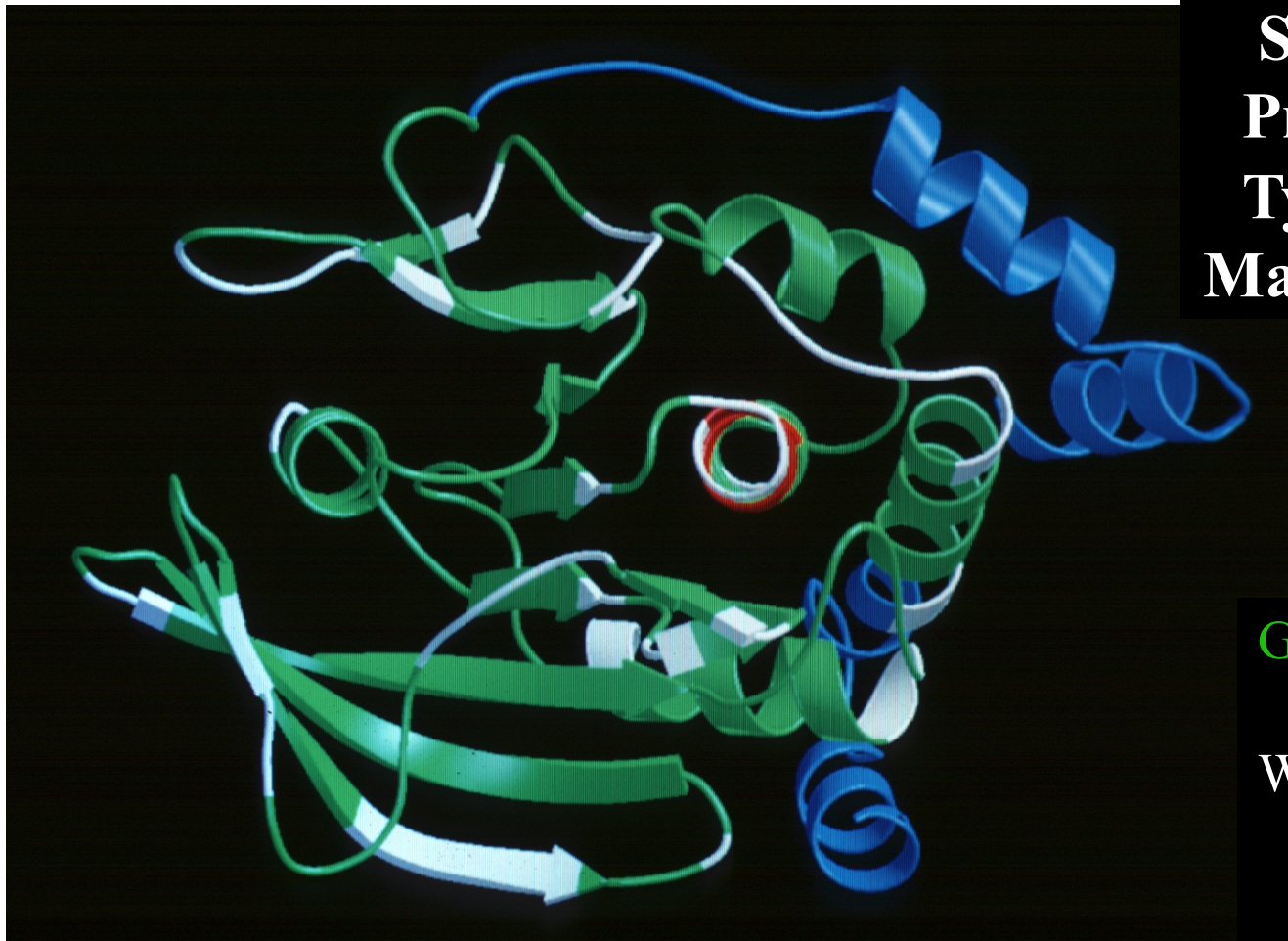
# Simple rules for secondary structure prediction by eye

Given an *accurate* multiple sequence alignment

- Positions of insertions and deletions - coil.
- Conserved Gly/Pro - turn.
- Run of conserved hydrophobics - buried  $\beta$ .
- Alternating conserved residues - surface  $\beta$ .
- 1,4,5,8 pattern of conserved residues -  $\alpha$ .
- 'pairs' of conserved residues -  $\alpha$ .



**Blind  
Secondary Structure  
Prediction for Protein  
Tyrosine Phosphatase  
Mapped on to Structure.**



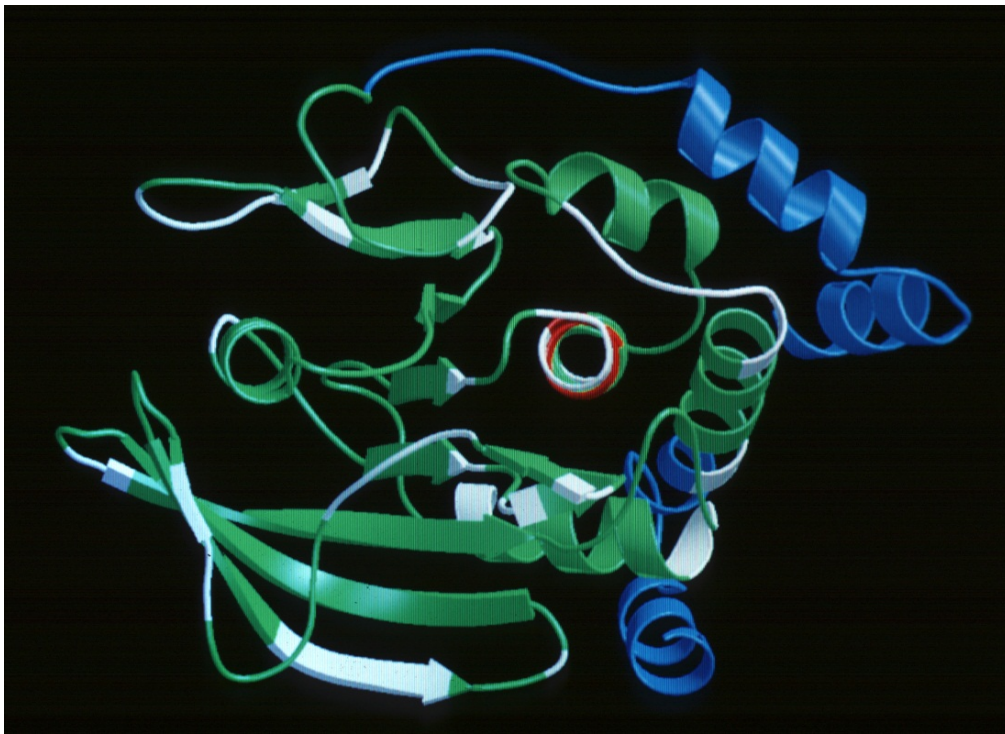
>75% Accuracy

Green: Correct Prediction.

White: Coil for strand/helix  
or Strand/helix for  
coil

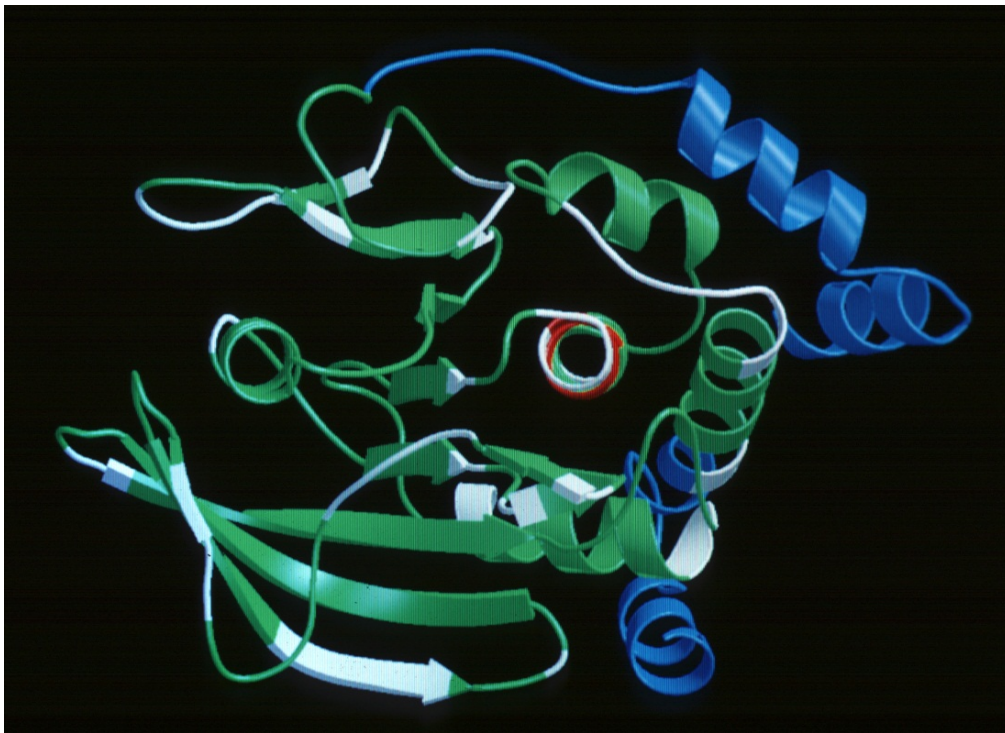
Red: Helix for strand or  
Strand for Helix

Blue: Not predicted.

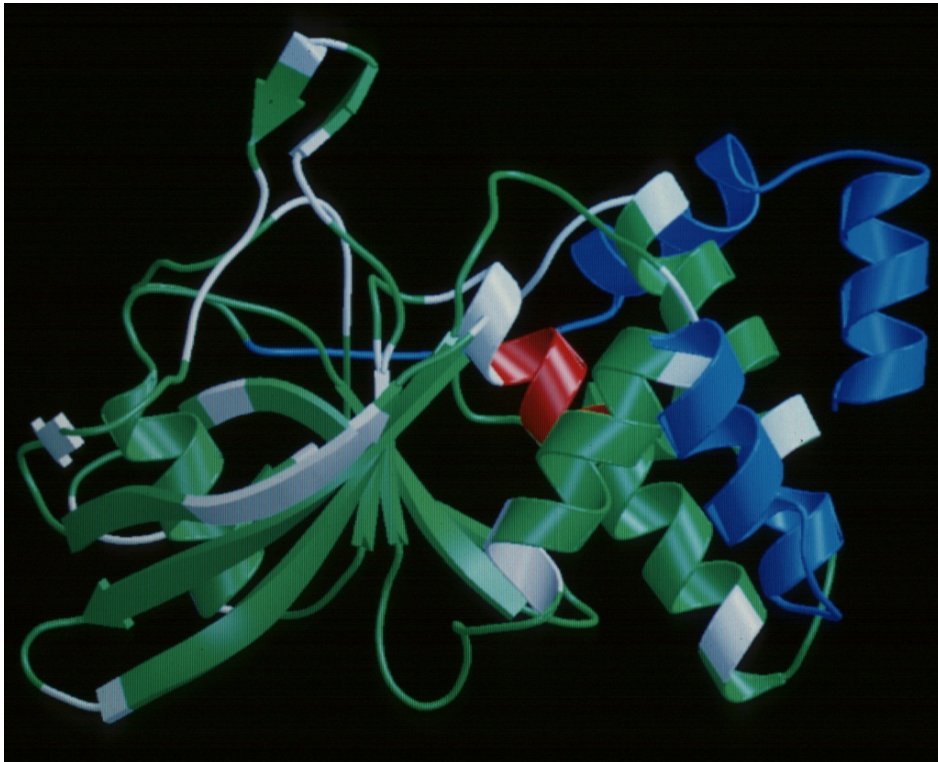


*Yersinia* Protein Tyrosine  
Phosphatase Prediction  
> 75% Accuracy





*Yersinia* Protein Tyrosine  
Phosphatase Prediction  
> 75% Accuracy



Oh Dear!

**End**