

# Tutorial on protein structure prediction

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# Outline of Talk

- 🦖 What is Bioengineering? Biomolecular Engineering? Bioinformatics?
- 🦖 What is a protein?
- 🦖 The folding problem and variants on it:
  - Local structure prediction
  - Fold recognition
  - Comparative modeling
  - “Ab initio” methods
  - Contact prediction
- 🦖 Protein Design



# What is Bioengineering?

Three concentrations:



## Biomolecular

- Drug design
- Biomolecular sensors
- Nanotechnology
- Bioinformatics



## Rehabilitation



## Bioelectronics



# What is Bioengineering?

Three concentrations:

 Biomolecular

 **Rehabilitation**

- Systems to help individuals with special needs
- Cell-phone-based systems to reach large numbers of people.
- Novel hardware to assist the blind
- Robotics for rehabilitation and surgery applications.

 Bioelectronics



# What is Bioengineering?

Three concentrations:

 Biomolecular

 Rehabilitation

 **Bioelectronics**

- Implantable devices
- Interfacing between organisms and electronics
- Artificial retina project



# What to take early

 Mathematics

 Chemistry and then biology

 Introductory bioengineering courses:

- BME80G, Bioethics (F)

- BME5, Intro to Biotechnology (W, S)

- CMPE80A: Universal Access: Disability, Technology, and Society (W, S)

 Declare your major immediately!!

You can always change to another one later.

Bioengineering is one of the most course-intensive majors on campus. Many courses have prerequisites. It's important to get advising office and faculty advice early.



# What is Biomolecular Engineering?

Engineering **with, of, or for** biomolecules. For example,

**with:** using proteins (or DNA, RNA, ...) as sensors or for self-assembly.

**of:** protein engineering—designing or artificially evolving proteins to have particular functions

**for:** designing high-throughput experimental methods to find out what molecules are present, how they are structured, and how they interact.



# What is Bioinformatics?

Bioinformatics: using computers and statistics to make sense out of the mountains of **data** produced by high-throughput experiments.

- 🦖 Genomics: annotating important sequences in genomes.
- 🦖 Phylogenetics: tree of life, ancestral genome reconstruction.
- 🦖 Systems biology: discovering and modeling biological networks.
- 🦖 Expression profiling: what genes are turned on under what conditions (DNA microarrays, RNAseq).
- 🦖 Protein structure and function prediction.
- 🦖 Proteomics: what proteins are present in a mixture.





# What is a protein?

- 🦋 There are many abstractions of a protein: a band on a gel, a string of letters, a mass spectrum, a set of 3D coordinates of atoms, a point in an interaction graph, . . . .
- 🦋 For us, a protein is a long skinny molecule (like a string of letter beads) that folds up consistently into a particular intricate shape.
- 🦋 The individual “beads” are amino acids, which have 6 atoms the same in each “bead” (the *backbone* atoms: N, H, CA, HA, C, O).
- 🦋 The final shape is different for different proteins and is essential to the function. The protein shapes are important, but are expensive to determine experimentally.



# Visualizing Proteins

There are many ways to look at proteins:

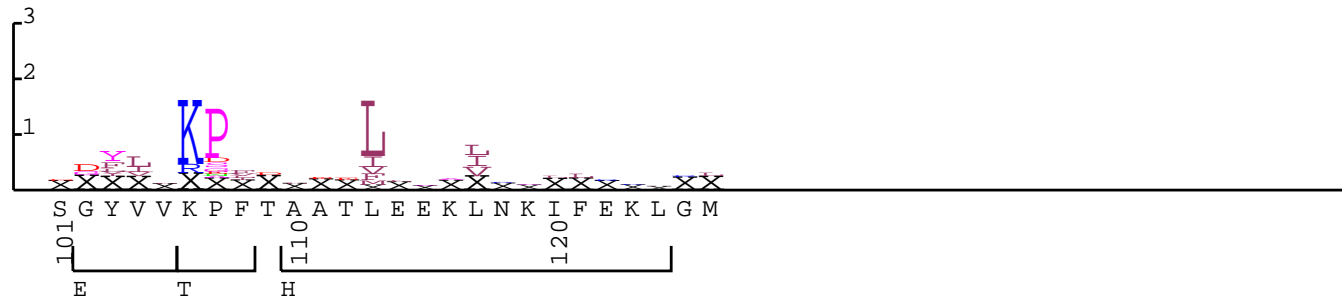
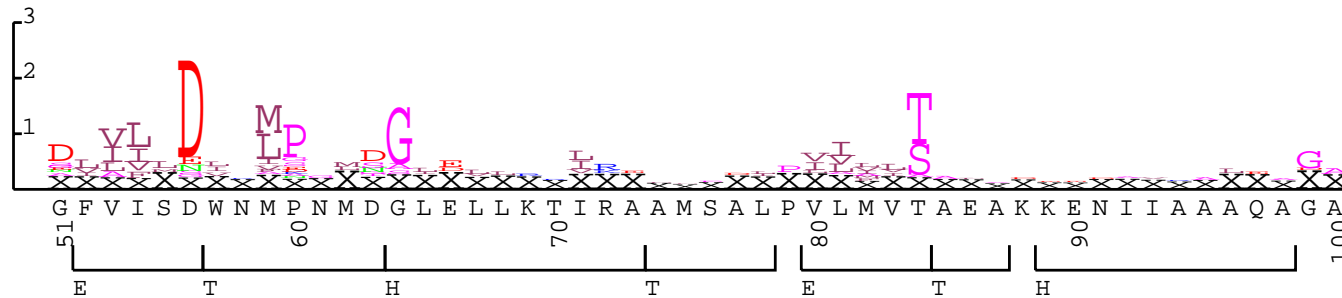
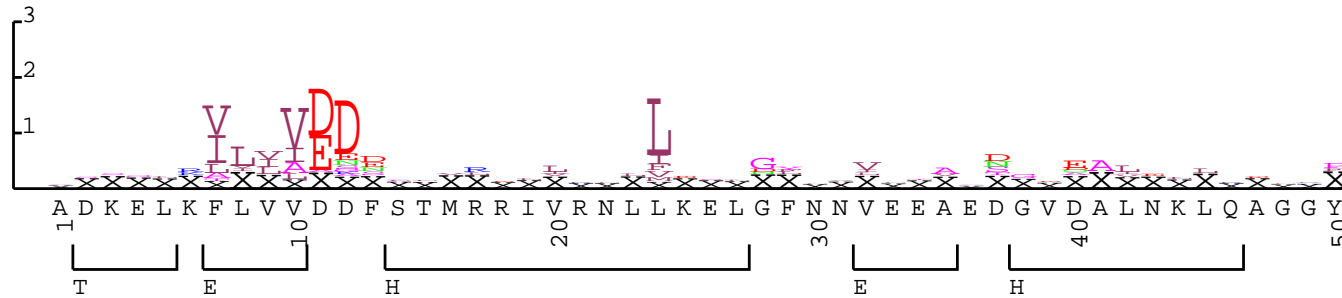
- 🦋 Strings of letters.
- 🦋 Sequence logos: letters plus conservation information.
- 🦋 Plastic models of structure.
- 🦋 Computer visualization of structure (rasmol, pymol, vmd, jmol, molmol, ... )



# Sequence logos (MSA)

Summarize multiple alignment for 1jbeA:

nostruct-align/1jbeA.t06 w0.5



# DEMO visualization

- 🦖 Demonstrate protein backbone using Darling Models
- 🦖 Demonstrate different views using Rasmol (or other viewer)



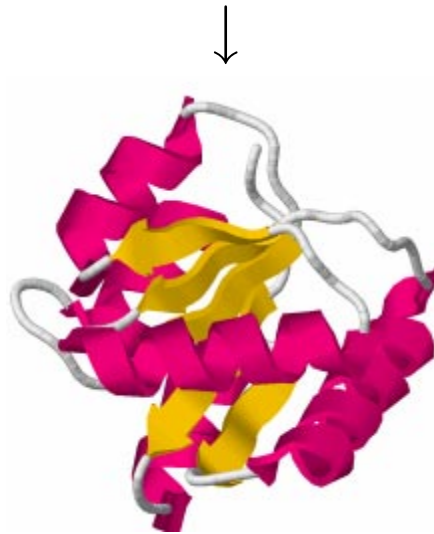
# Folding Problem

The *Folding Problem*:

If we are given a sequence of amino acids (the letters on a string of beads), can we predict how it folds up in 3-space?

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```
>1jbeA Chemotaxis protein CHEY from E. coli  
ADKELKFLVVDDFSTMRRIVRNLLKELGFNNVEEAEDGVDALNKLQAGGY  
GFVISDWNMPNMDGLELLKTIRADGAMSALPVLMTAEAKKENIIAAAQA  
GASGYVVKPFTAATLEEKLNKIFEKLG M
```



Too hard!



# Fold-recognition problem

The *Fold-recognition Problem*:

Given a sequence of amino acids  $A$  (the *target* sequence) and a library of proteins with known 3-D structures (the *template* library), figure out which templates  $A$  match best, and align the target to the templates.

- 🦖 The backbone for the target sequence is predicted to be very similar to the backbone of the chosen template.



# New-fold prediction

- ⚠️ What if there is *no* template we can use?
- ⚠️ We can try to generate many conformations of the protein backbone and try to recognize the most protein-like of them.
- ⚠️ Search space is huge, so we need a good conformation generator and a cheap cost function to evaluate conformations.



# Secondary structure Prediction

- 🦖 Instead of predicting the entire structure, we can predict local properties of the structure.
- 🦖 What local properties do we choose?
- 🦖 We want properties that are well-conserved through evolution, easily predicted, and useful for finding and aligning templates.
- 🦖 One popular choice is a 3-valued helix/strand/other alphabet—we have investigated many others. Typically, predictors get about 80% accuracy on 3-state prediction.
- 🦖 Many machine-learning methods have been applied to this problem, but the most successful are neural networks.





# Contact prediction

- 🚧 Try to predict which residues come close to each other.
- 🚧 Ones close along the chain are easy (secondary structure prediction).
- 🚧 Ones far apart along chain, but close in space, are hard to predict, but most useful.
- 🚧 Correlated mutation is powerful indication of close residues.



# (Rational) Protein Design

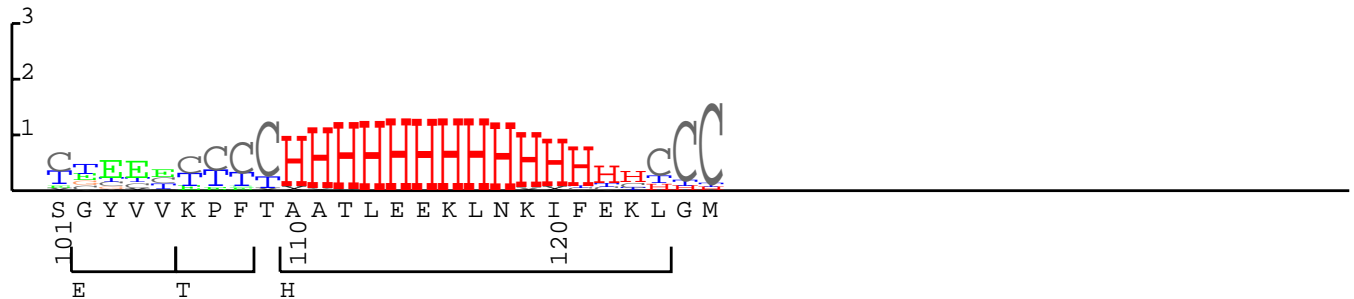
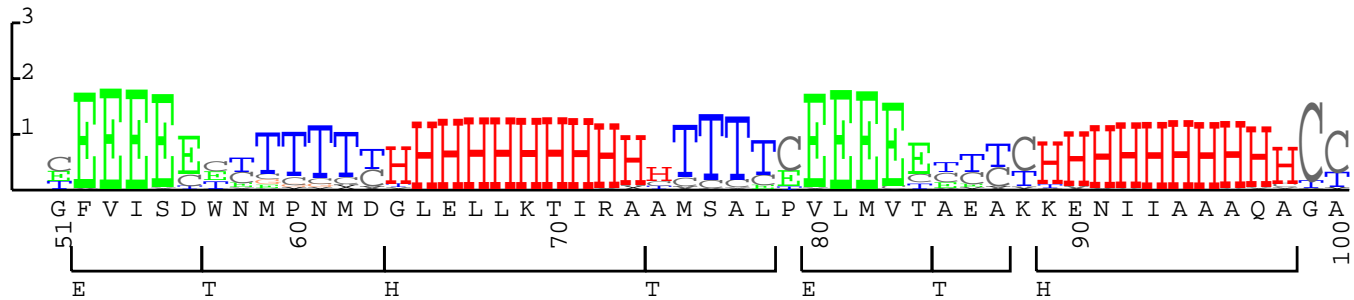
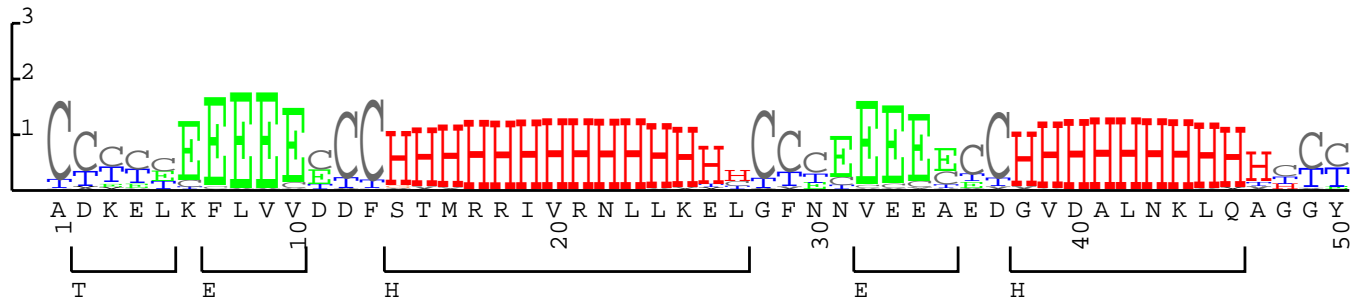
- 🦖 New direction for Karplus lab.
- 🦖 Use neural nets to predict amino acids from local structure properties.
- 🦖 Use Undertaker to build models.
- 🦖 Use RosettaDesign (from Baker lab) to modify sequences.
- 🦖 Use Undertaker, Rosetta, and Gromacs to validate that designed structure is good.
- 🦖 Target applications: short proteins that mimic agouti-related protein (and other proteins that bind melanocortin receptors) but which do not have disulfide bridges.



# Sequence logos (NN)

Summarize local structure prediction:

nostruct-align/1jbeA.t06 EBGHTL



# CASP Competition Experiment

- 🦖 Everything published in literature “works”
- 🦖 CASP set up as true blind test of prediction methods.
- 🦖 Sequences of proteins about to be solved released to prediction community.
- 🦖 Predictions registered with organizers.
- 🦖 Experimental structures compared with solution by assessors.
- 🦖 “Winners” get papers in *Proteins: Structure, Function, and Bioinformatics*.



# Overview of Prediction Method

- 🦖 Look for homologs.
  - Homologs = proteins with common ancestral sequence.
  - Can't really determine algorithmically, so we look for "sufficiently similar" sequences.
- 🦖 Make multiple sequence alignment (MSA).



# Overview of Prediction Method 2

- 🦖 Use MSA to make local structure predictions.
- 🦖 Use MSA (and local structure predictions) to make Hidden Markov Models (HMMs).
- 🦖 Use HMMs to find and align proteins of known structure (PDB).
- 🦖 Use model-building program to change alignments into 3D models.
- 🦖 Clean up models (close gaps, rebuild loops, adjust sidechains, ...)
- 🦖 Choose best model(s) (Model Quality Assessment).
- 🦖 Maybe use contact predictions to select among models.

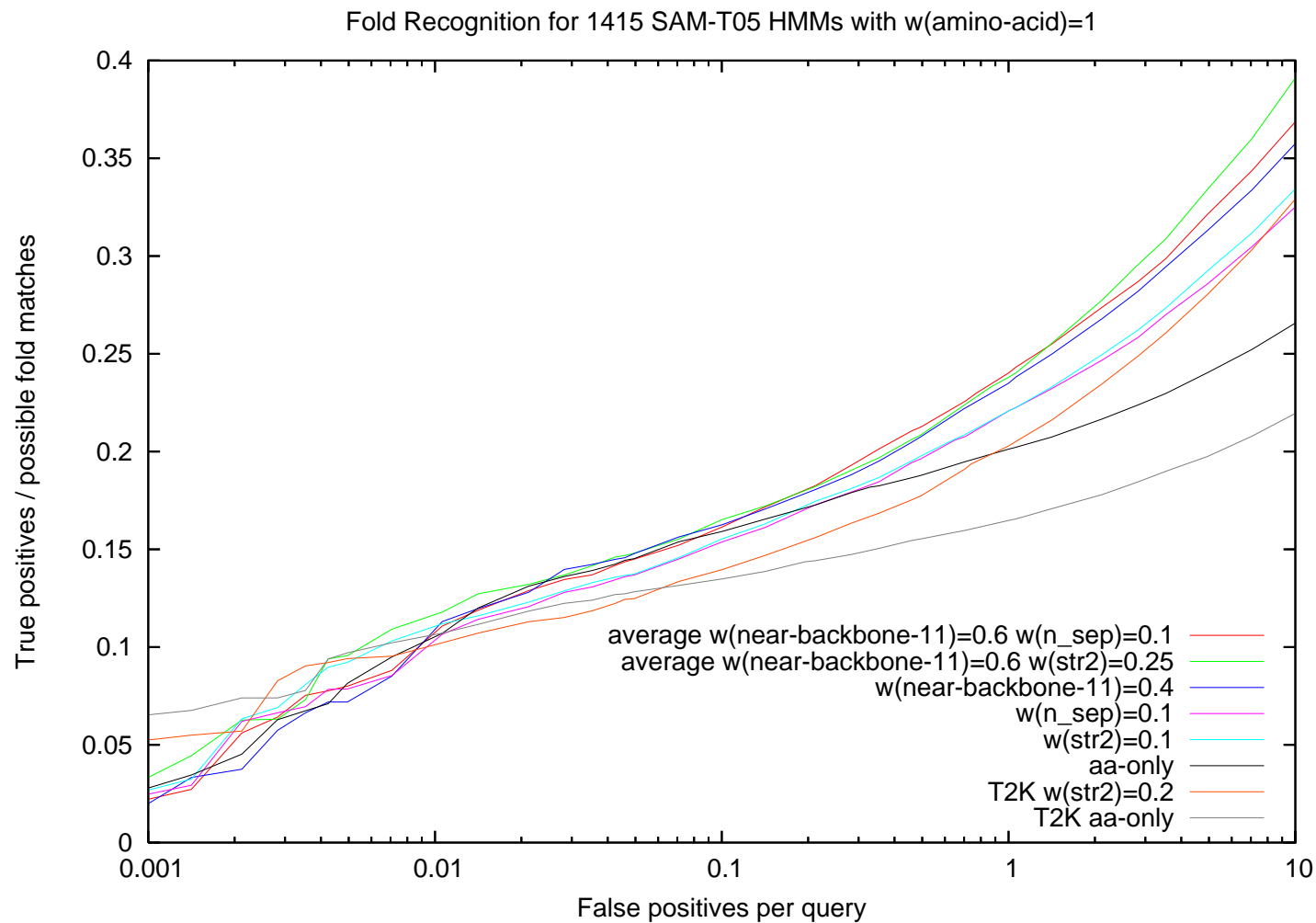


# Contact Prediction Method

- 🦖 Use mutual information between columns.
- 🦖 Thin alignments aggressively (30%, 35%, 40%, 50%, 62%).
- 🦖 Compute e-value for mutual info (correcting for small-sample effects).
- 🦖 Compute rank of  $\log(\text{e-value})$  within protein.
- 🦖 Feed  $\log(\text{e-values})$ , log rank, contact potential, joint entropy, and separation along chain for pair, and amino-acid profile, predicted burial, and predicted secondary structure for each residue of pair into a neural net.



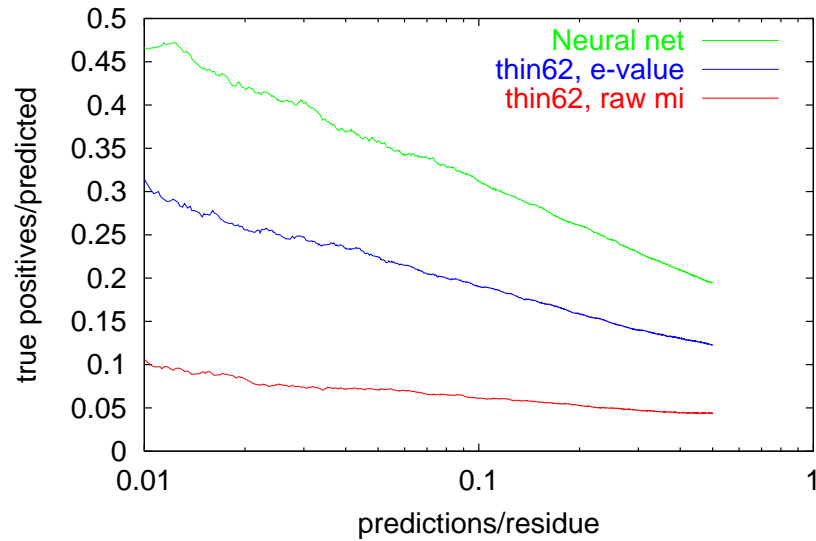
# Fold recognition results



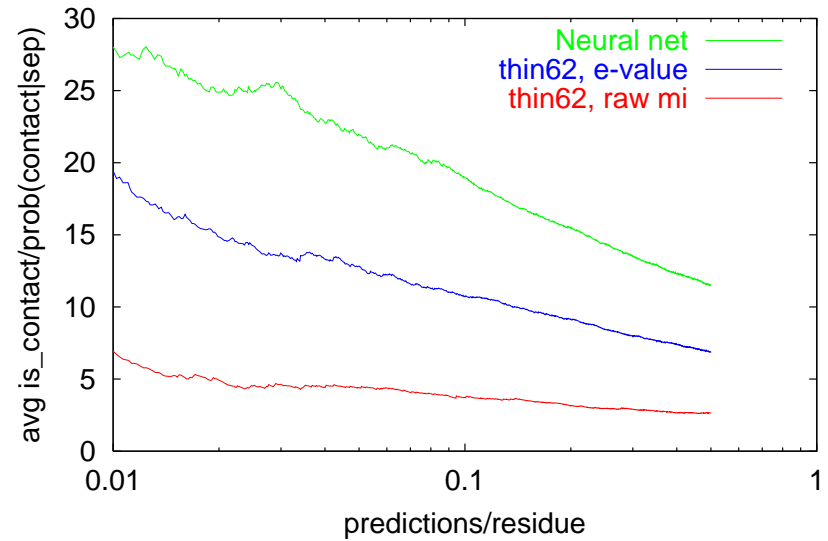


# Contact prediction results

Accuracy of contact prediction, by protein

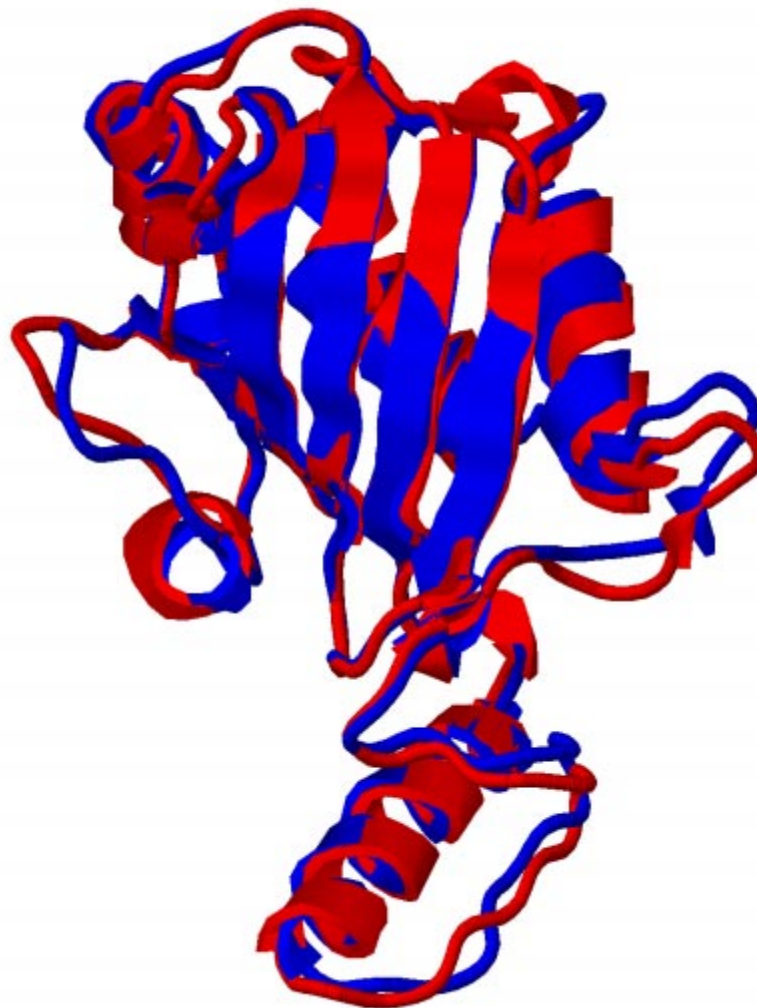


Weighted-accuracy of contact prediction, by protein



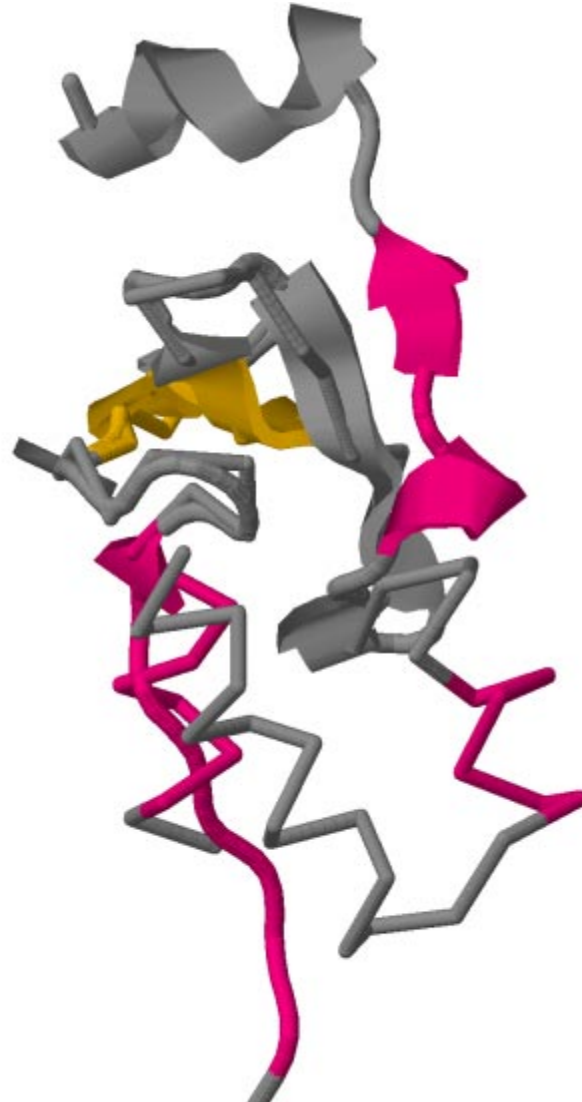
# T0298 domain 2 (130–315)

RMSD= 2.468Å all-atom, 1.7567Å  $C_{\alpha}$ , GDT=82.5%  
best model 1 submitted to CASP7 (red=real)



# Comparative modeling: T0348

RMSD= 11.8 Å  $C_{\alpha}$ , GDT=58.2% (cartoon=real)  
best model 1 by CASP7 GDT, Robetta1 slightly better.

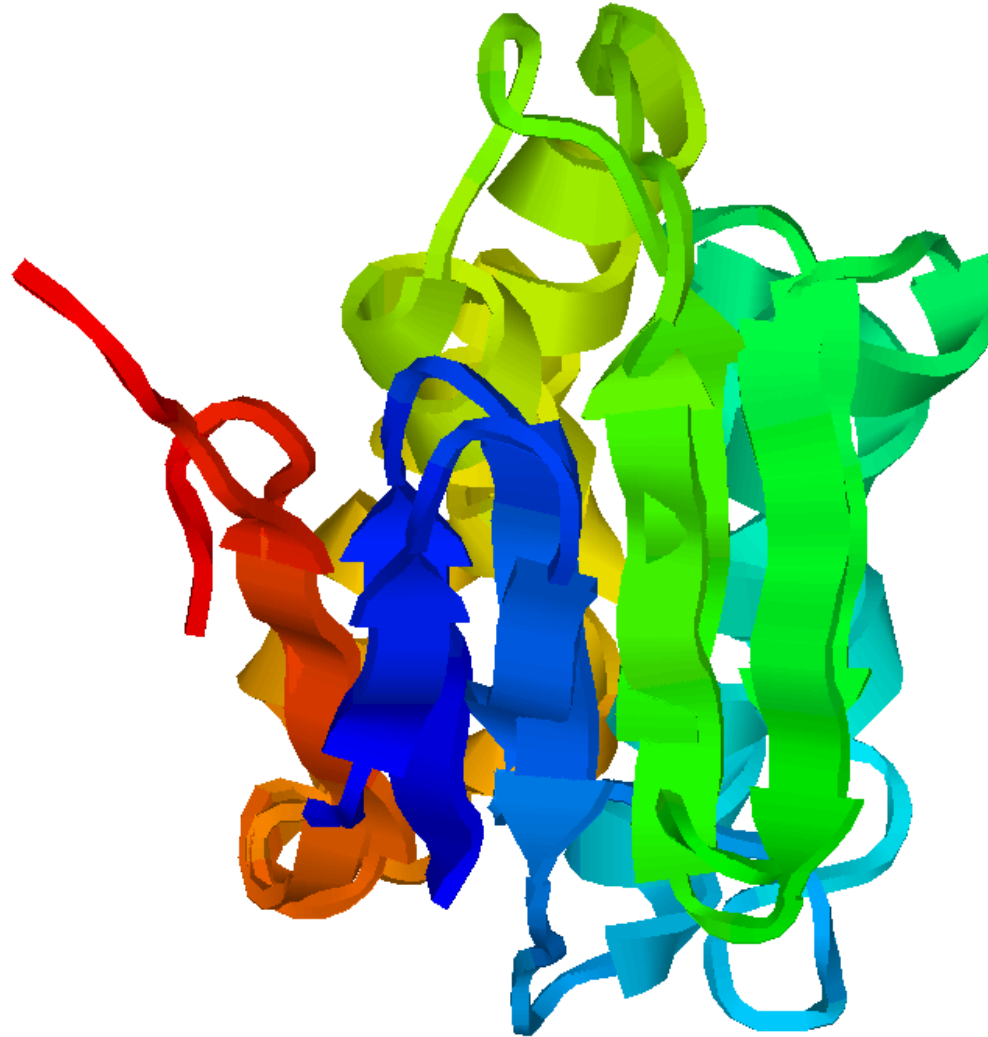


# Target T0201 (NF, CASP6)

- 🚩 We tried forcing various sheet topologies and selected 4 by hand.
- 🚩 Model 1 has right topology (5.912Å all-atom, 5.219Å  $C_{\alpha}$ ).
- 🚩 Unconstrained cost function not good at choosing topology (two strands curled into helices).
- 🚩 Helices were too short.



# Target T0201 (NF, CASP6)

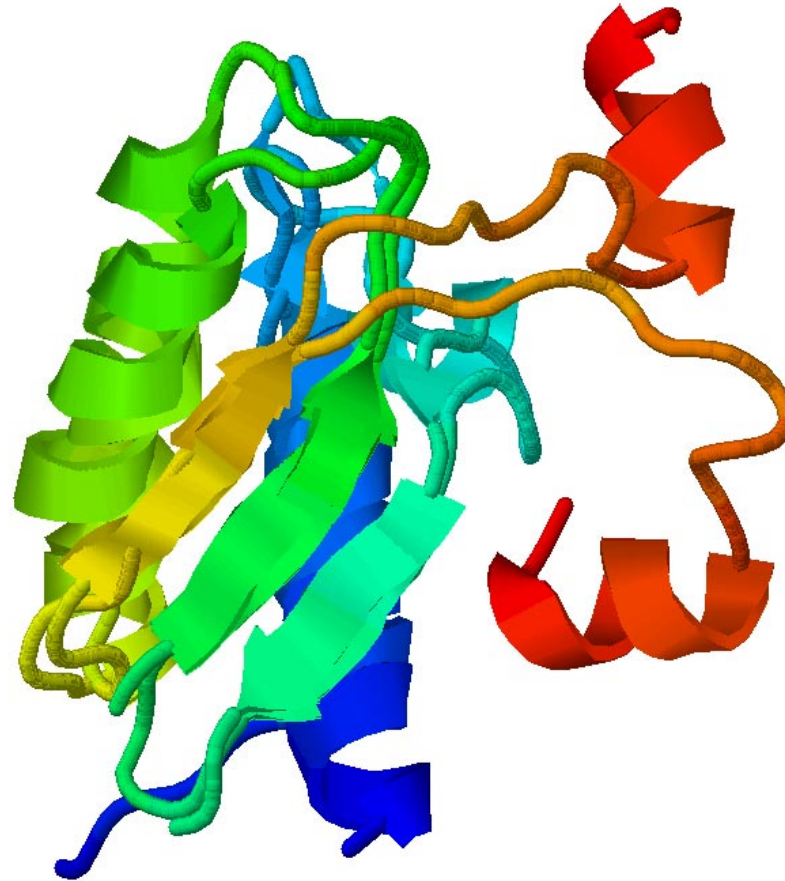


# Target T0230 (FR/A, CASP6)

- 🦖 Good except for C-terminal loop and helix flopped wrong way.
- 🦖 We have secondary structure right, including phase of beta strands.
- 🦖 Contact prediction helped, but we put too much weight on it—decoys fit predictions better than real structure does.

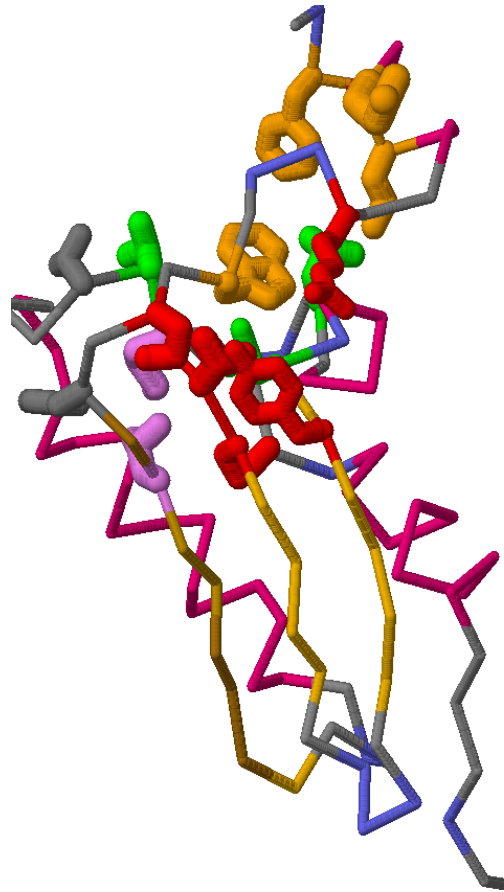


# Target T0230 (FR/A, CASP6)



# Target T0230 (FR/A)

Real structure with contact predictions:





# Web sites

**These slides:** <http://www.soe.ucsc.edu/~karplus/papers/structure-prediction-tutorial-jul-2009.pdf>

**Old CASP results—all our results and working notes:**

<http://www.soe.ucsc.edu/~karplus/casp6/>

<http://www.soe.ucsc.edu/~karplus/casp7/>

<http://www.soe.ucsc.edu/~karplus/casp8/>

**SAM-T08 prediction server:**

[http://compbio.soe.ucsc.edu/SAM\\_T08/T08-query.html](http://compbio.soe.ucsc.edu/SAM_T08/T08-query.html)

**UCSC bioinformatics and bioengineering degree programs:**

<http://www.bme.ucsc.edu/bioinformatics/>

<http://beng.soe.ucsc.edu/>

