Karplus lab: protein structure prediction and design

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Outline of Talk

- & What is a protein?
- A The folding problem and variants on it:
 - Local structure prediction
 - Fold recognition
 - Comparative modeling
 - "Ab initio" methods
 - Contact prediction
- 💪 Protein Design



What is a protein?

- A 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.
- A The individual "beads" are amino acids, which have 6 atoms the same in each "bead" (the *backbone* atoms: N, H, CA, HA, C, O).
- A The final shape is different for different proteins and is essential to the function.

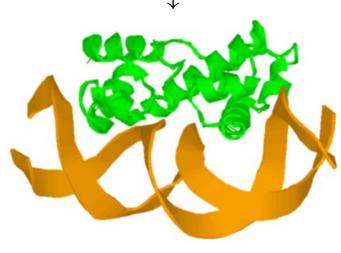


The protein shapes are important, but are expensive to determine experimentally.

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?

MTMSRRNTDA ITIHSILDWI EDNLESPLSL EKVSERSGYS KWHLQRMFKK ETGHSLGQYI RSRKMTEIAQ KLKESNEPIL YLAERYGFES QQTLTRTFKN YFDVPPHKYR MTNMQGESRF LHPLNHYNS



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.

A 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.
- Generation And Antices Antices
- Many machine-learning methods have been applied to this problem, but the most successful is neural networks.
- Using Conditional Random Fields can improve sampling of sequences, without improving accuracy on individual residues.



Local Structure Alphabets

- & What local properties do we choose?
- We want properties that are well-conserved through evolution, easily predicted, and useful for finding and aligning templates.
- & We have investigated many alphabets.
- Current favorites are str2, a 13-state secondary-structure alphabet that distinguishes between different β strands, and near-backbone-11, and 11-state burial alphabet.



Contact prediction

- A Predict that residues separated along the chain are clos in 3-space.
- **4** Use mutual information between columns.
- 4 Thin alignments aggressively (30%, 35%, 40%, 50%, 62%).
- Compute e-value for mutual info (correcting for small-sample effects).
- Compute rank of log(e-value) within protein.
- Feed log(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.



(Rational) Protein Design

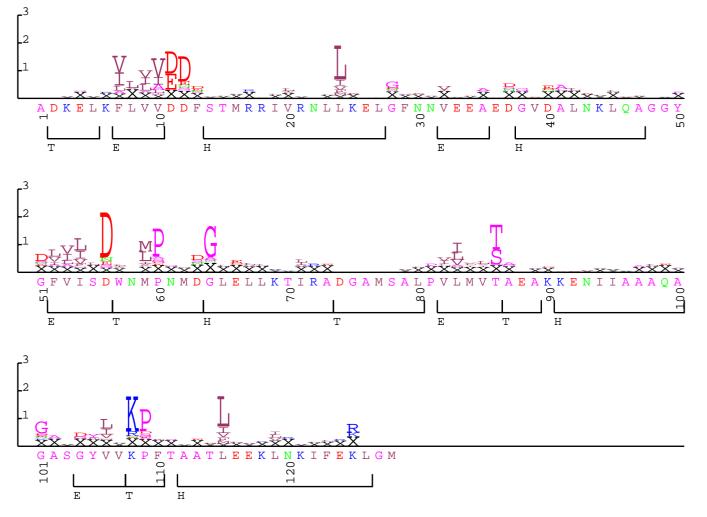
- A New direction for lab.
- Use local-structure neural nets in reverse (find sequences highly predicted to have right local structure).
- Train new neural nets to take local-structure inputs and provide amino-acid outputs.
- Use undertaker to build models.
- Use RosettaDesign to modify sequences.
- Target applications: specific binding of carbon nanotubes, mimics for AGRP (agouti-related protein) binding to different melanocortin receptors.



Sequence logos (MSA)

Summarize multiple alignment:

nostruct-align/3chy.t2k w0.5

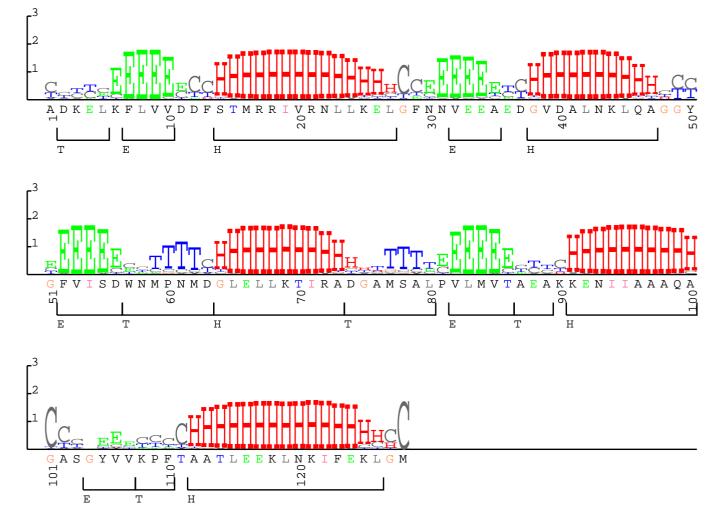




Sequence logos (NN)

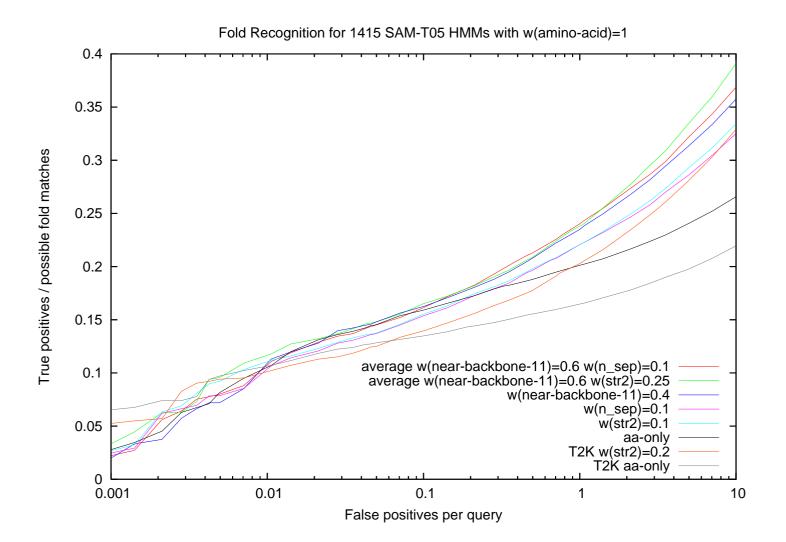
Summarize local structure prediction:

nostruct-align/3chy.t2k EBGHTL



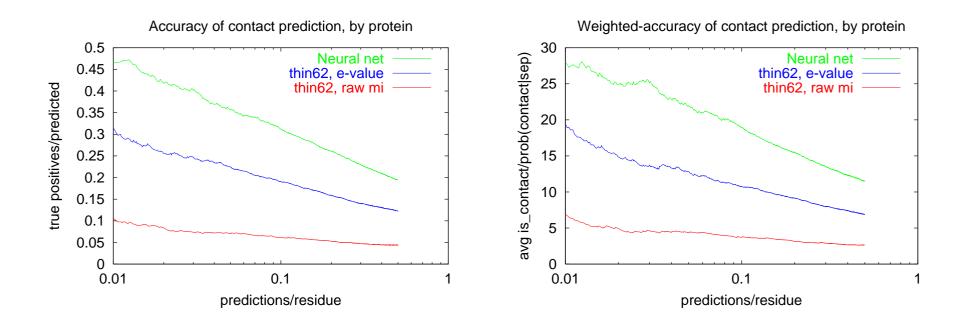


Fold recognition results





Contact prediction results





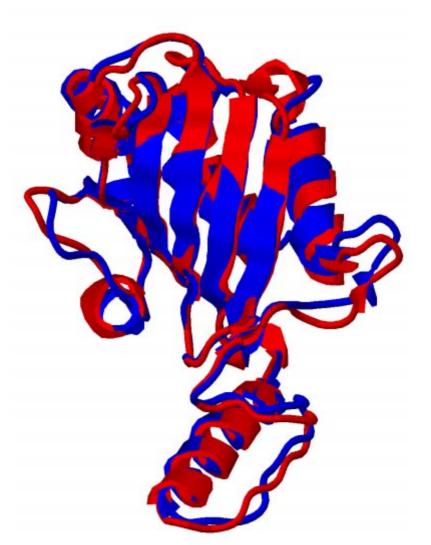
CASP Competition Experiment

- 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.
- A Predictions registered with organizers.
- Experimental structures compared with solution by assessors.
- Structure, Function, and Bioinformatics.



T0298 domain 2 (130–315)

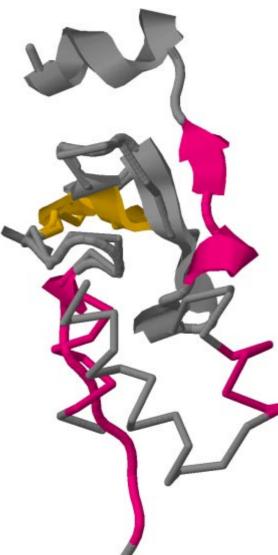
RMSD= 2.468Å all-atom, 1.7567Å C_{α} , GDT=82.5% best model 1 submitted to CASP7 (red=real)





Comparative modeling: T0348

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



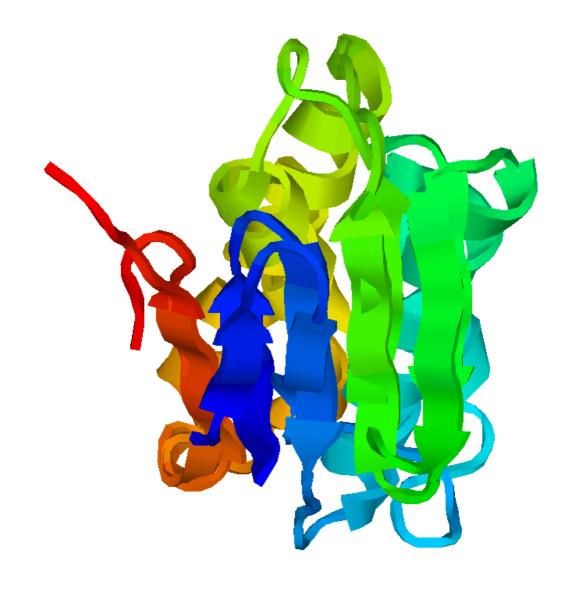


Target T0201 (NF, CASP6)

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



Target T0201 (NF, CASP6)



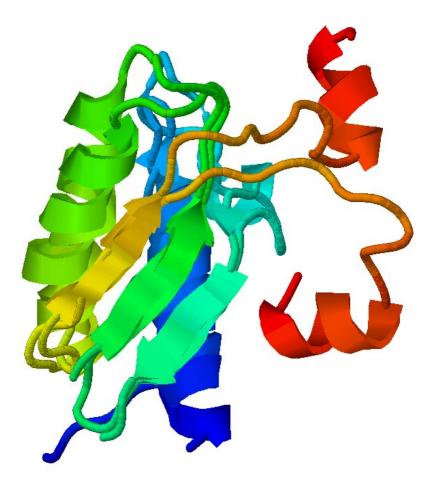


Target T0230 (FR/A, CASP6)

- 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.



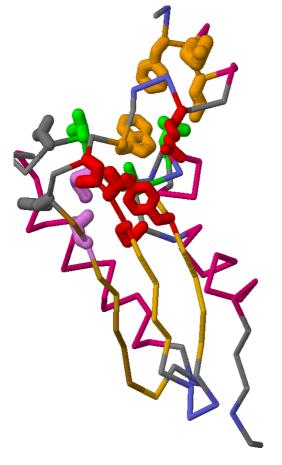
Target T0230 (FR/A, CASP6)





Target T0230 (FR/A)

Real structure with contact predictions:





Web sites

These slides:

http://users.soe.ucsc.edu/~karplus/papers/what-lab-does-oct-2009.pdf

CASP2 through CASP8—all our results and working notes:

http://users.soe.ucsc.edu/~karplus/casp2/

. . .

http://users.soe.ucsc.edu/~karplus/casp8/

SAM-T08 prediction server:

http://compbio.soe.ucsc.edu/SAM_T08/T08-query.html

UCSC bioinformatics (research and degree programs) info:

http://compbio.soe.ucsc.edu/

