

# Secondary Protein Structure Determination

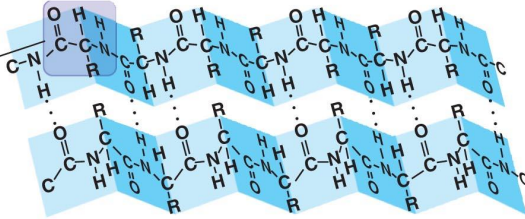
Big Data Final Project  
Fall 2016  
Rachael Kretsch



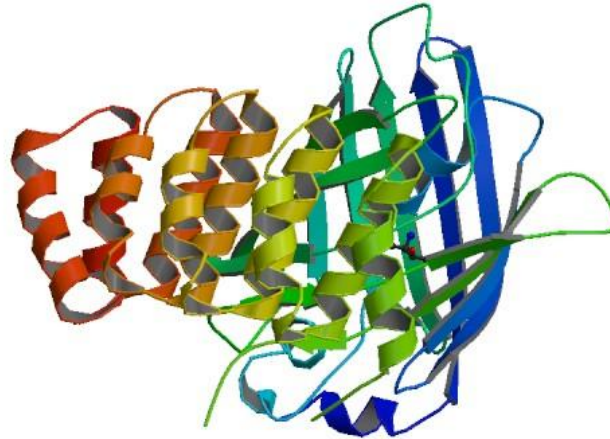
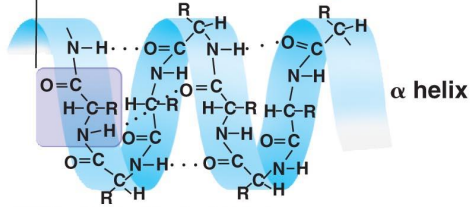
# Protein Structure

## Secondary Structure

$\beta$  pleated sheet



Examples of amino acid subunits



Amino acids

**Primary protein structure**  
sequence of a chain of amino acids

Pleated sheet

Alpha helix

**Secondary protein structure**  
hydrogen bonding of the peptide backbone causes the amino acids to fold into a repeating pattern

**Tertiary protein structure**  
three-dimensional folding pattern of a protein due to side chain interactions

**Quaternary protein structure**  
protein consisting of more than one amino acid chain

# Chromoproteins and Fluorescent Proteins

“Chromoprotein” and “Fluorescent” PDB

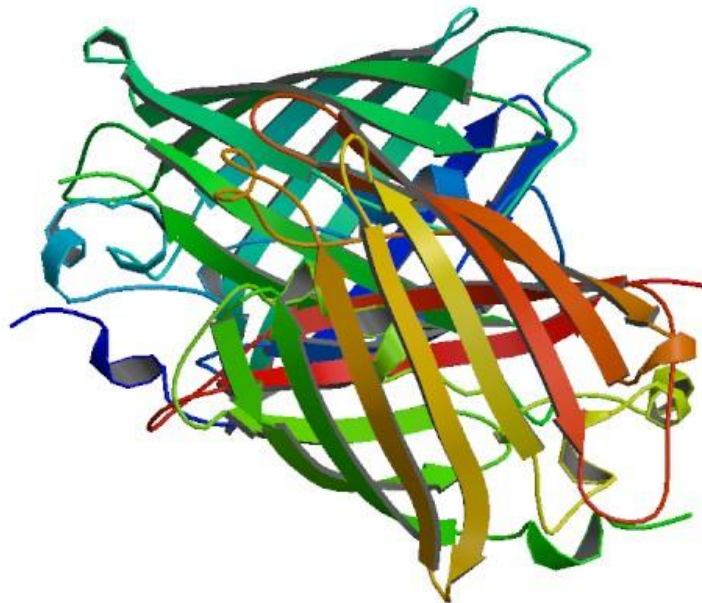
5 data points removed <25

Only took one form of each protein

739 total proteins

392,106 amino acids

2,744,742 values



SOPM 239 proteins (1994) , 267 proteins GORIV (1996), s2d 2671 proteins (2014)

# Data Base

124,928 proteins



# My algorithm

Interpret fasta

Ignore unusual amino acids  
Ignore duplicates of same sequence ID  
Delete short sequences (<25)

Get structure

Dssp: standardized secondary structure assignment library for PDB from NMR and/or crystallography data

Score

Helix (1): alpha-helices, 3-helices  
Coil (0): turns, coils, bends, 5-helices  
Beta (1): strands, bridges

Get matrices

7 scores per amino acid

Logistic regression

L2, reg=3.4

Assess

Accuracy: Q3

$$Q3 = \sum_{n=1}^k NC(i)/NO(i)$$

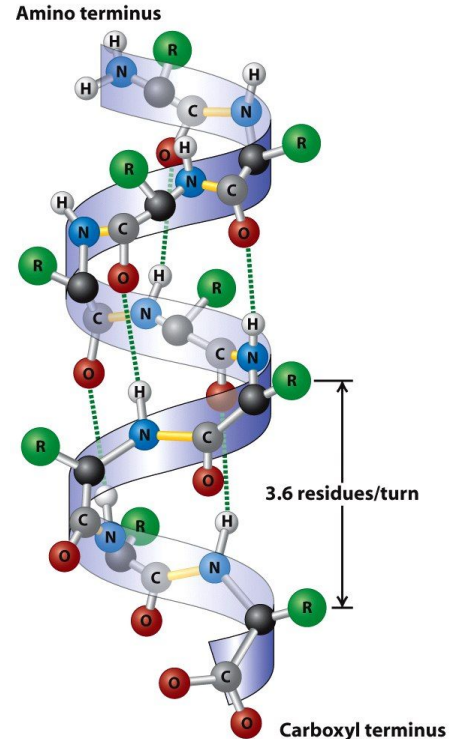
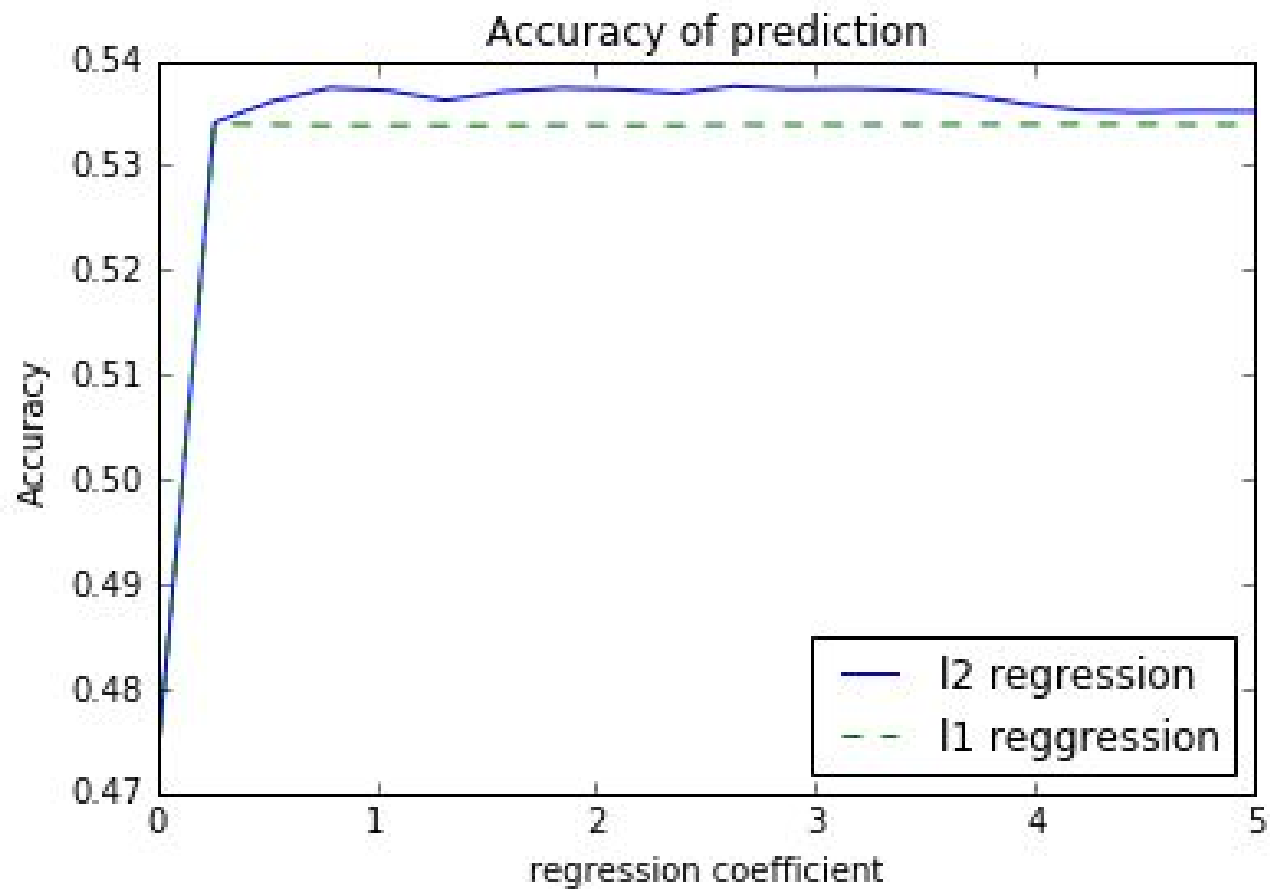
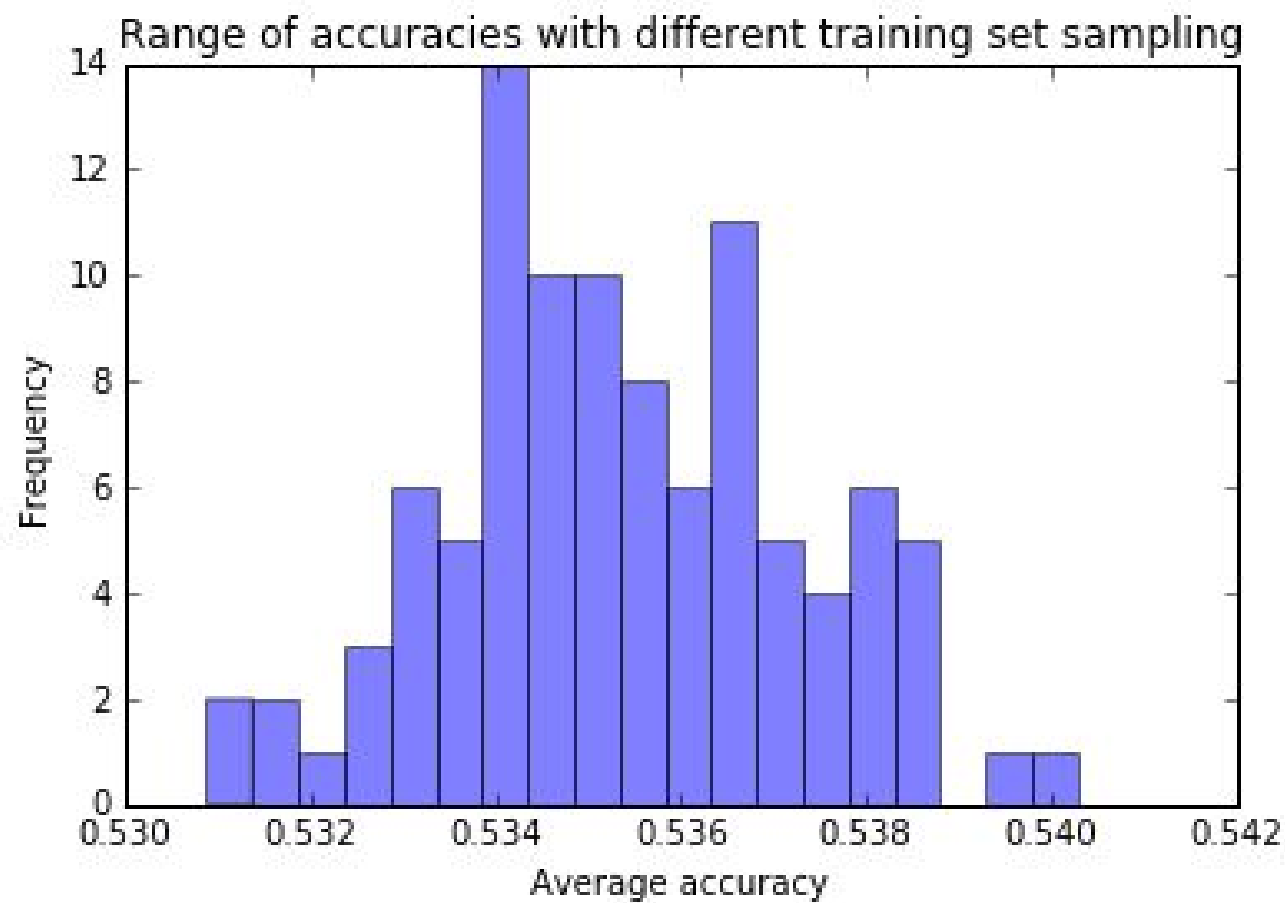
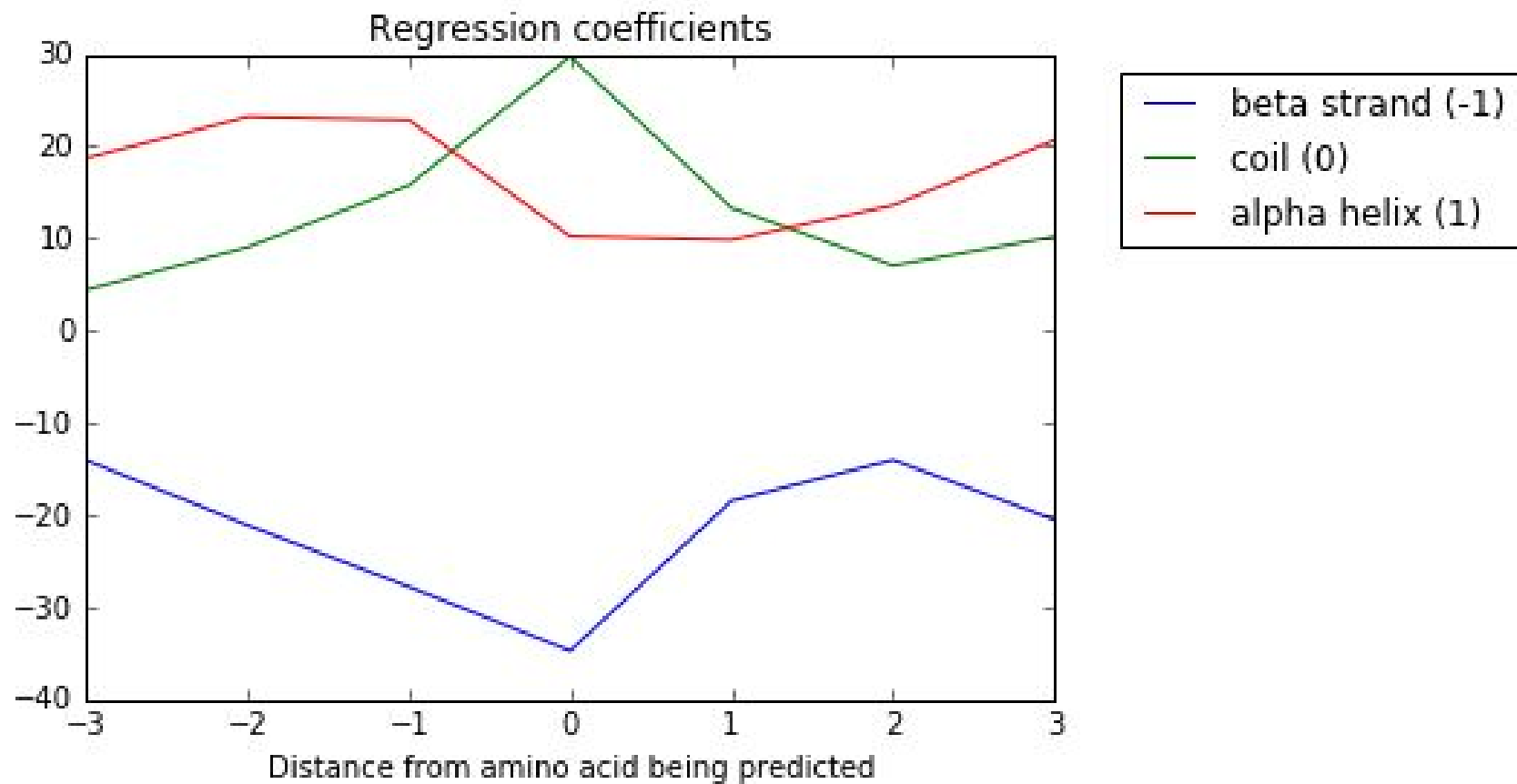


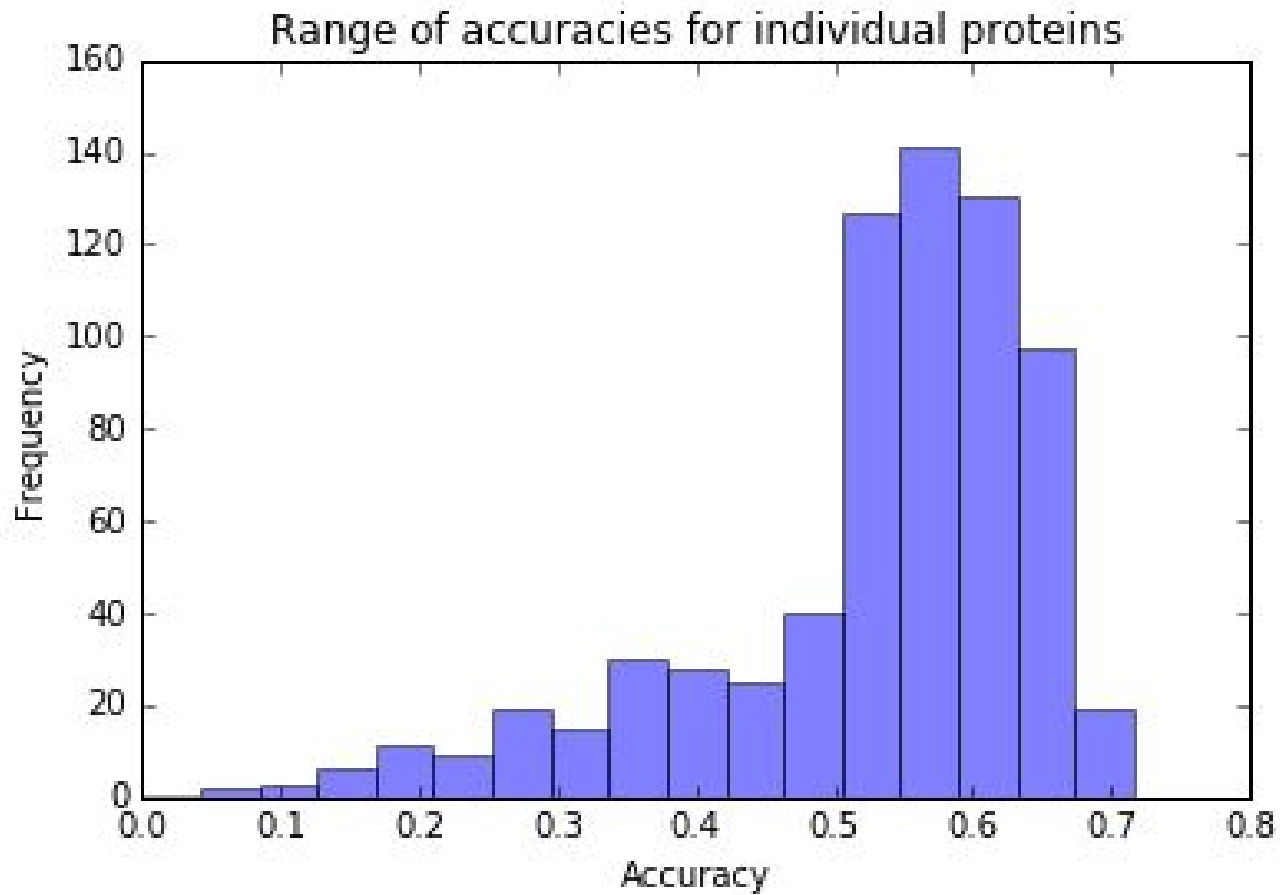
Figure 3-4  
Molecular Cell Biology, Sixth Edition  
© 2008 W. H. Freeman and Company



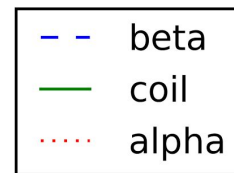
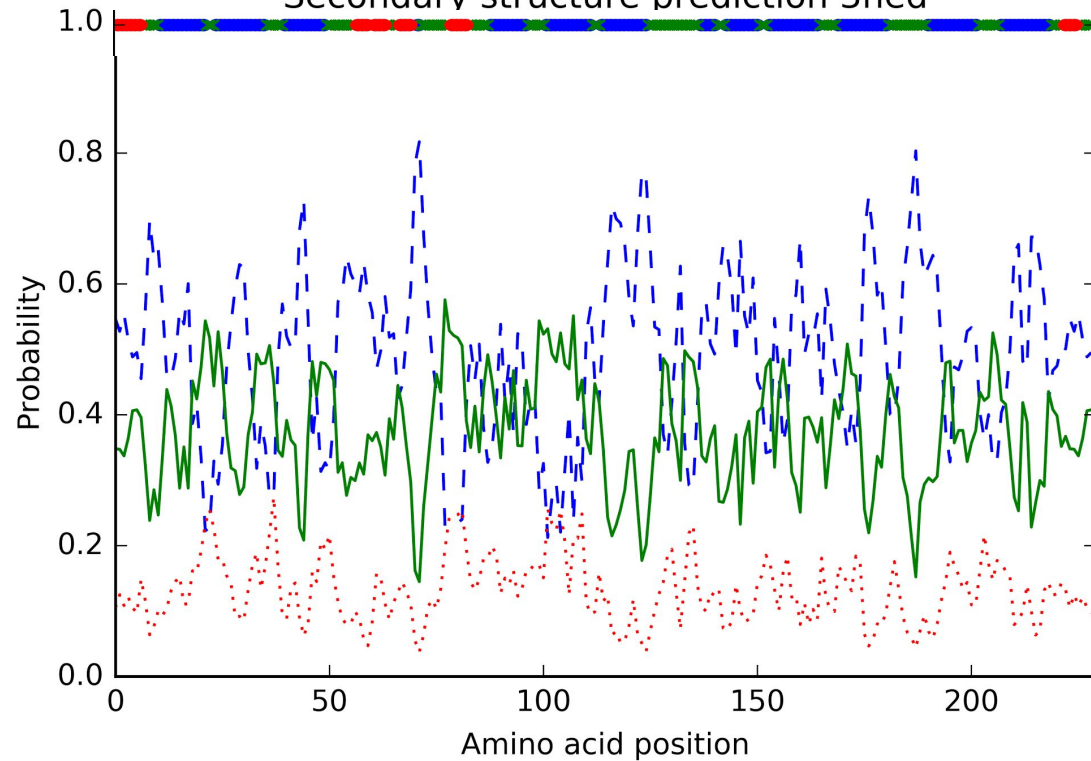




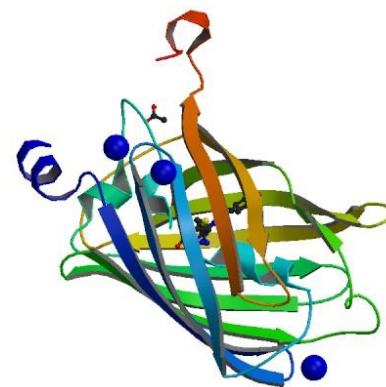




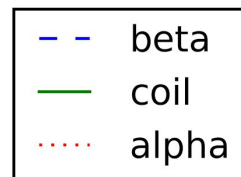
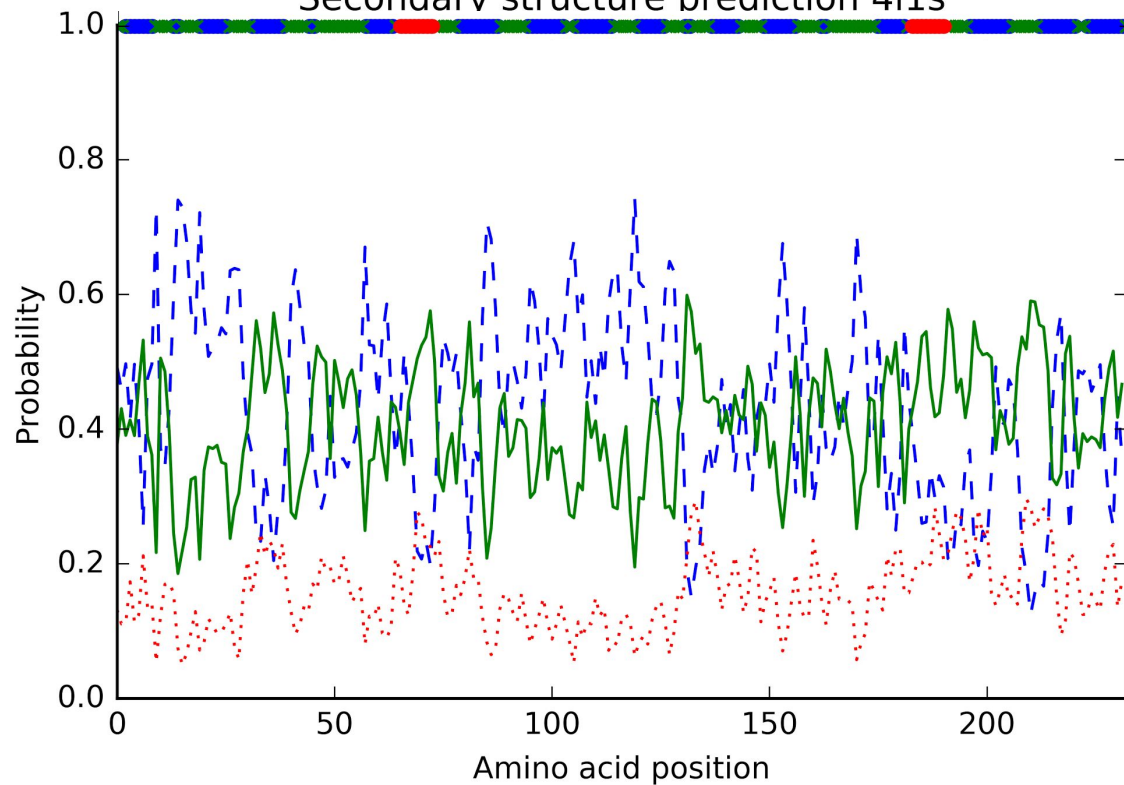
### Secondary structure prediction 3ned



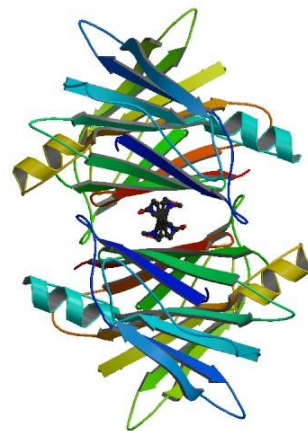
accuracy = 0.5983



### Secondary structure prediction 411s



accuracy = 0.2629



# GOR IV

Information Theory  
Bayesian Statistics

Scoring algorithm

$$I(\Delta S_i; R_1, \dots, R_n) = \log\left[\frac{P(S_i, R_1, \dots, R_n)}{P(n - S_i, R_1, \dots, R_n)}\right] + \log\left[\frac{P(n - S)}{P(S)}\right]$$

$$\begin{aligned} \log\left[\frac{P(S_i, R_1, \dots, R_{17})}{P(n - S_i, R_1, \dots, R_{17})}\right] &= \frac{2}{17} \sum_{m=-8, n>m}^{+8} \log\left[\frac{P(S_i, R_i + m, R_i + n)}{P(n - S_i, R_i + m, R_i + m)}\right] \\ &\quad - \frac{15}{17} \sum_{m=-8}^{+8} \log\left[\frac{P(S_i, R_i + m)}{P(n - S_i, R_i + m)}\right] \end{aligned}$$

# SOPM



→ 7 amino acid frames

Input sequence

Build a sub databases

Optimization

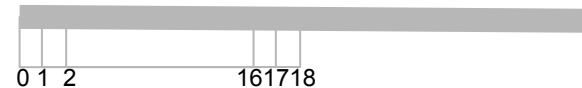
Prediction

For all proteins in sub database:  
Predict secondary structure by  
Sequence homology with other proteins

$$f_k(i+1) = f(i) + \frac{NO(i) - NP_k(i)}{NP_k(i)}$$

$$Q3 = \sum_{n=1}^k NC(i)/NO(i)$$

Predict input



s2D

BMRB dataset

s2D dataset

PSI-BLAST

Data sets

B-heavy data set

SLFN (window=11)

SLFN (window=15)

Secondary-structure populations

N-to-1 network

Mean secondary-structure pop

SLFN

Corrections (window=5)

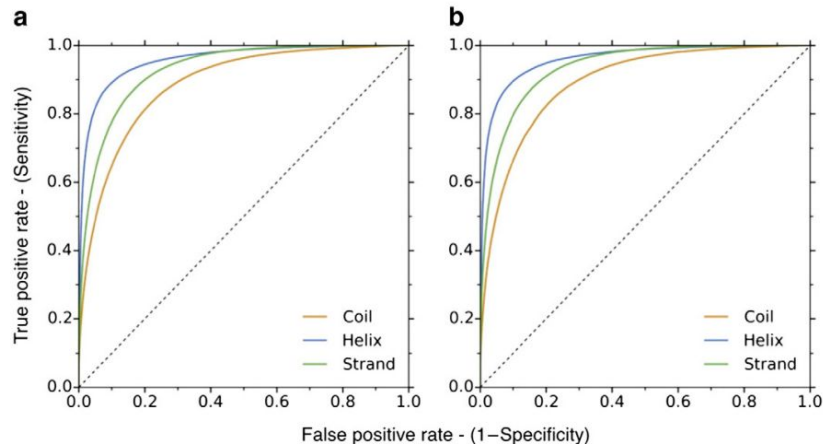
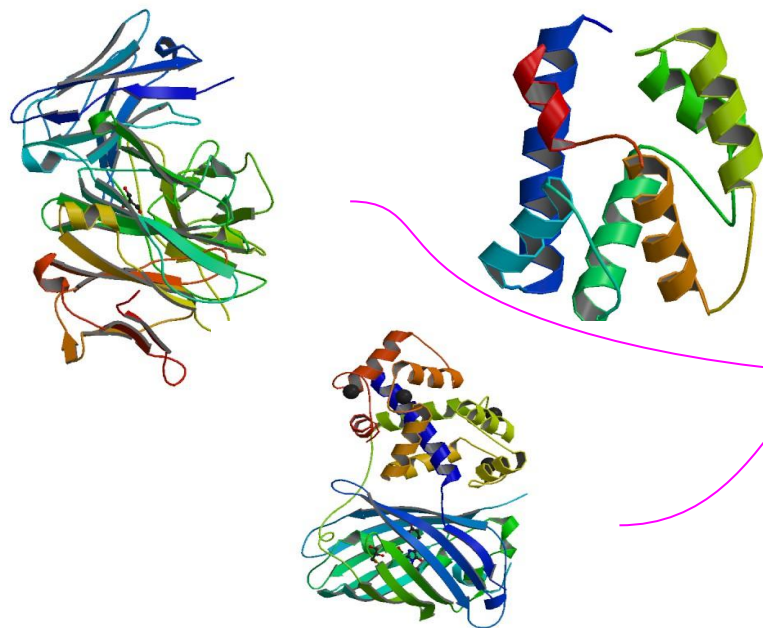


Table 1: Accuracy of various secondary structure prediction methods

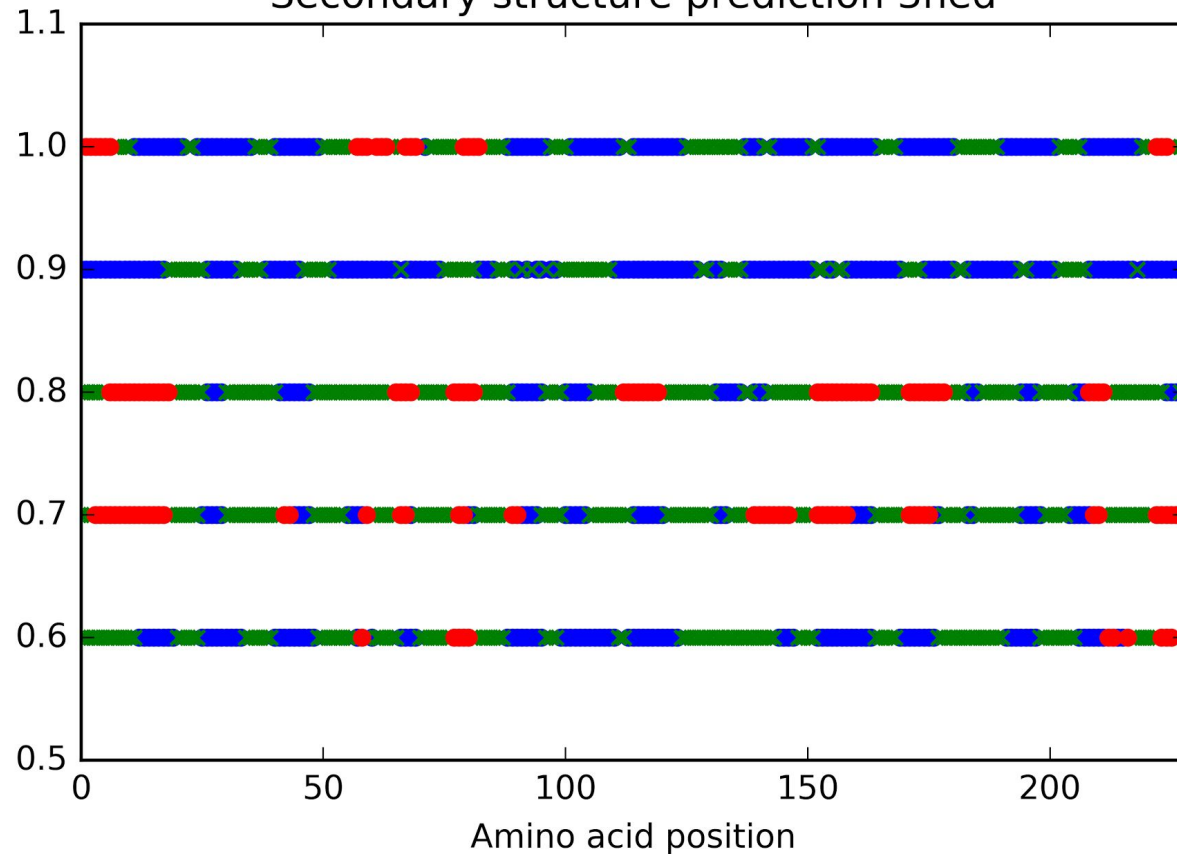
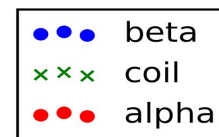
Method	Accuracy on data set (%)	Reported accuracy (%)
Logistic regression	54	-
GORIV	48	64
SOPM	51	69
s2D	64	85-88

Table 2: Accuracy of methods for specific proteins

Protein ID	Accuracies (%)			
	Logistic regression	GORIV	SOPM	s2D
1bgp	25	64	62	64
4q7t	25	44	42	67
4qgw	25	69	61	83
5h88	26	37	37	57
4l1s	26	64	52	70
5h89	27	37	39	60
3s0f	27	49	58	70
4q9w	27	51	55	70
3rwt	27	37	35	48
5hzo	28	37	46	64
1bfp	60	48	60	77
3ekh	60	54	61	55
3ned	60	40	46	73
4k3g	60	49	54	59
3cfc	60	58	60	61
1xkh	60	50	48	47
2wht	60	37	58	70
4w6b	60	44	54	69
4xvp	60	44	49	52
3dqh	60	42	56	73



## Secondary structure prediction 3ned



Correct structure

Logistic regression  
Accuracy = 60%

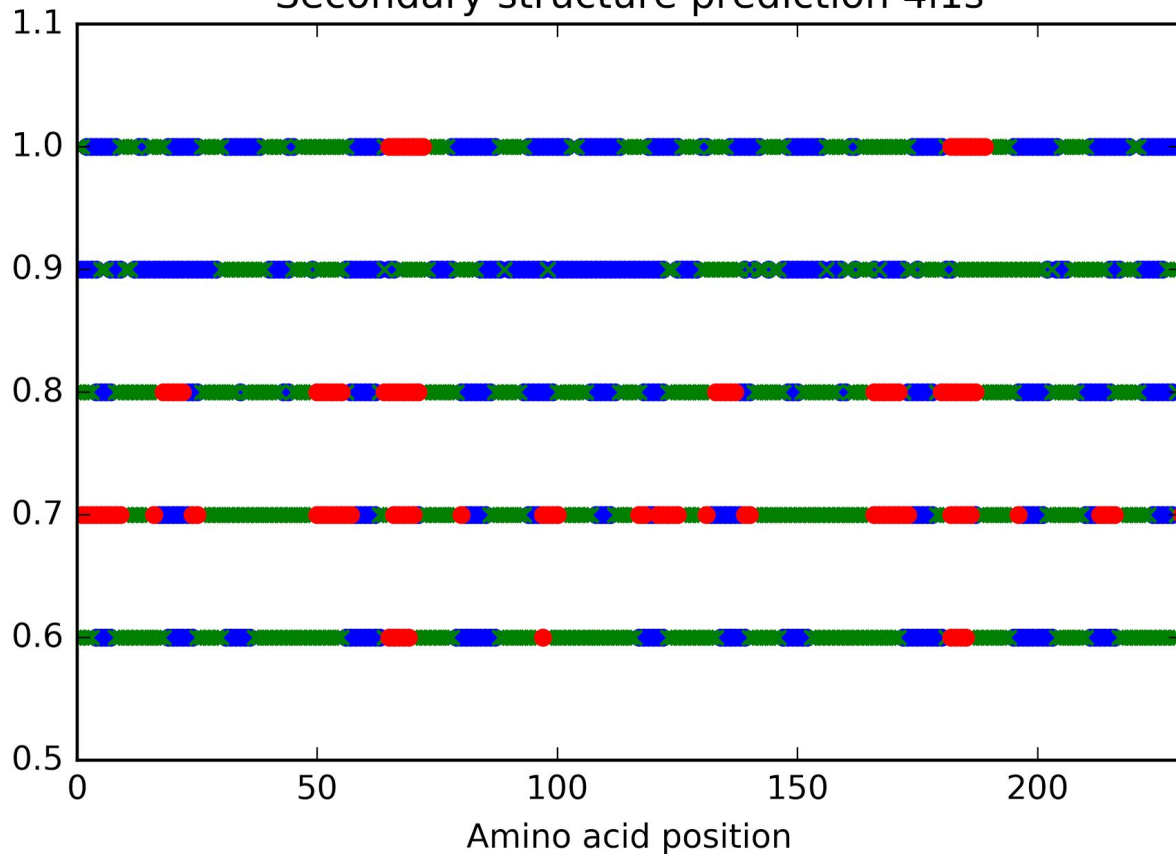
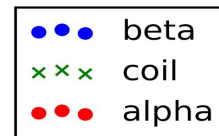
GORIV  
Accuracy = 40%

SOPM  
Accuracy = 46%

s2D  
Accuracy = 73%



## Secondary structure prediction 4l1s



Correct structure

Logistic regression  
Accuracy = 26%

GORIV  
Accuracy = 64%

SOPM  
Accuracy = 52%

s2D  
Accuracy = 70%

# Conclusions

For a group of proteins with conserved structures, training on similar proteins is beneficial.

Bias of training set affects generalization.

Helices may require longer range interactions.

Logistic regression on a biased data set did outperform older prediction methods, but neural network methods are still more accurate.

# Future Directions

Protein folding

Post transcriptional modifications

Promising techniques with bias training sets

Other methods: HCAM, YASSP

Directed mutations in the lab

# Literature Cited

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# Questions?

