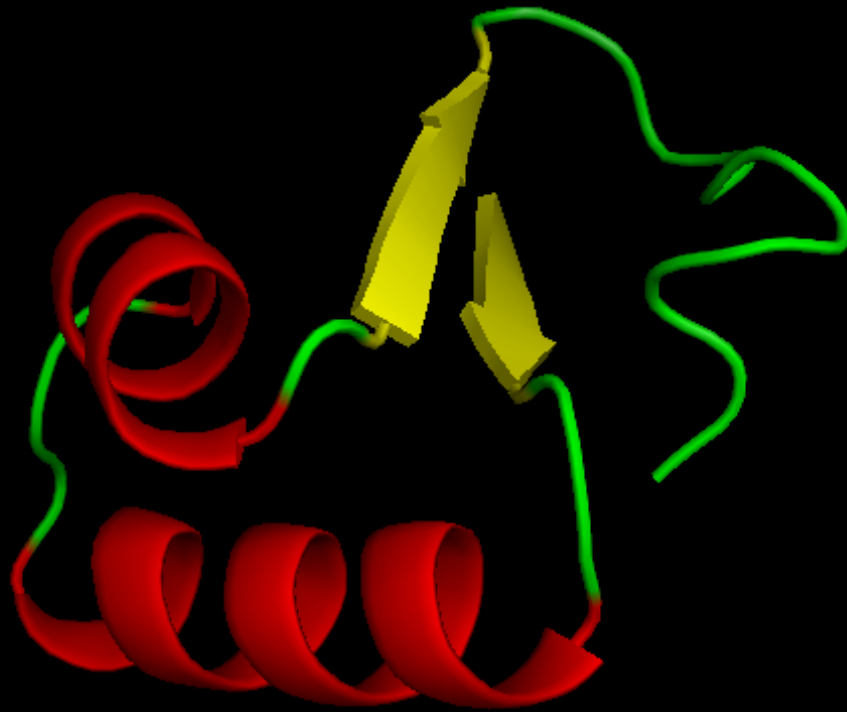


Structural biology

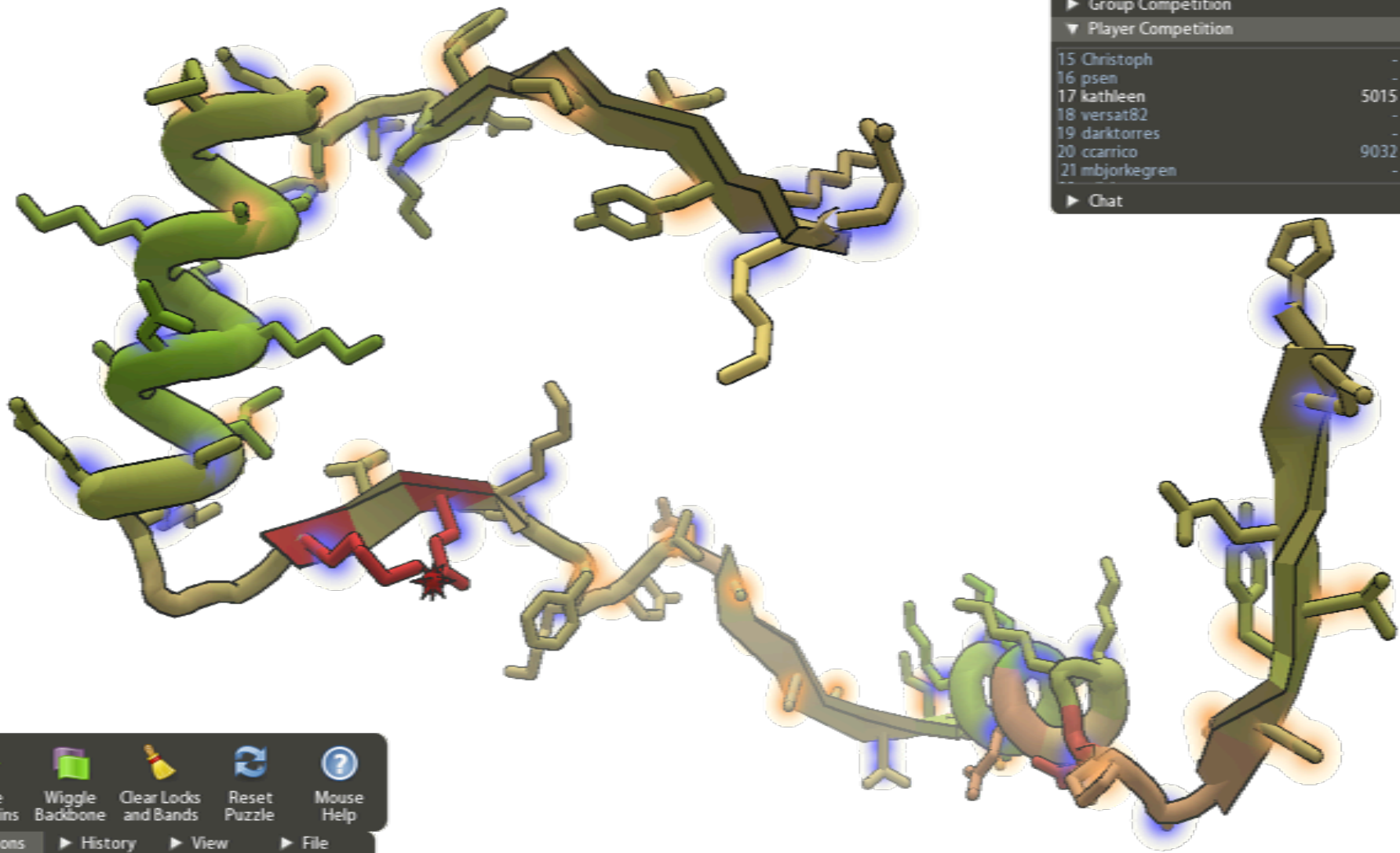
From sequence to structure

Proteins form into distinct shapes



Aims of structural biology

- Predict the confirmation of a poly-peptide chain
- Predict and analyze the function of a protein








Rank: 17 Score: 5015

48: Pro Peptide

- ▶ Group Competition
- ▼ Player Competition

15	Christoph	-	9101
16	pseu	-	9098
17	kathleen	5015	9092
18	versat82	-	9091
19	darktorres	-	9081
20	ccarrico	9032	9066
21	mbjorkegren	-	9048

▶ Chat

 Shake Sidechains
  Wiggle Backbone
  Clear Locks and Bands
  Reset Puzzle
  Mouse Help

▲ Actions ▶ History ▶ View ▶ File

 Pull Tool

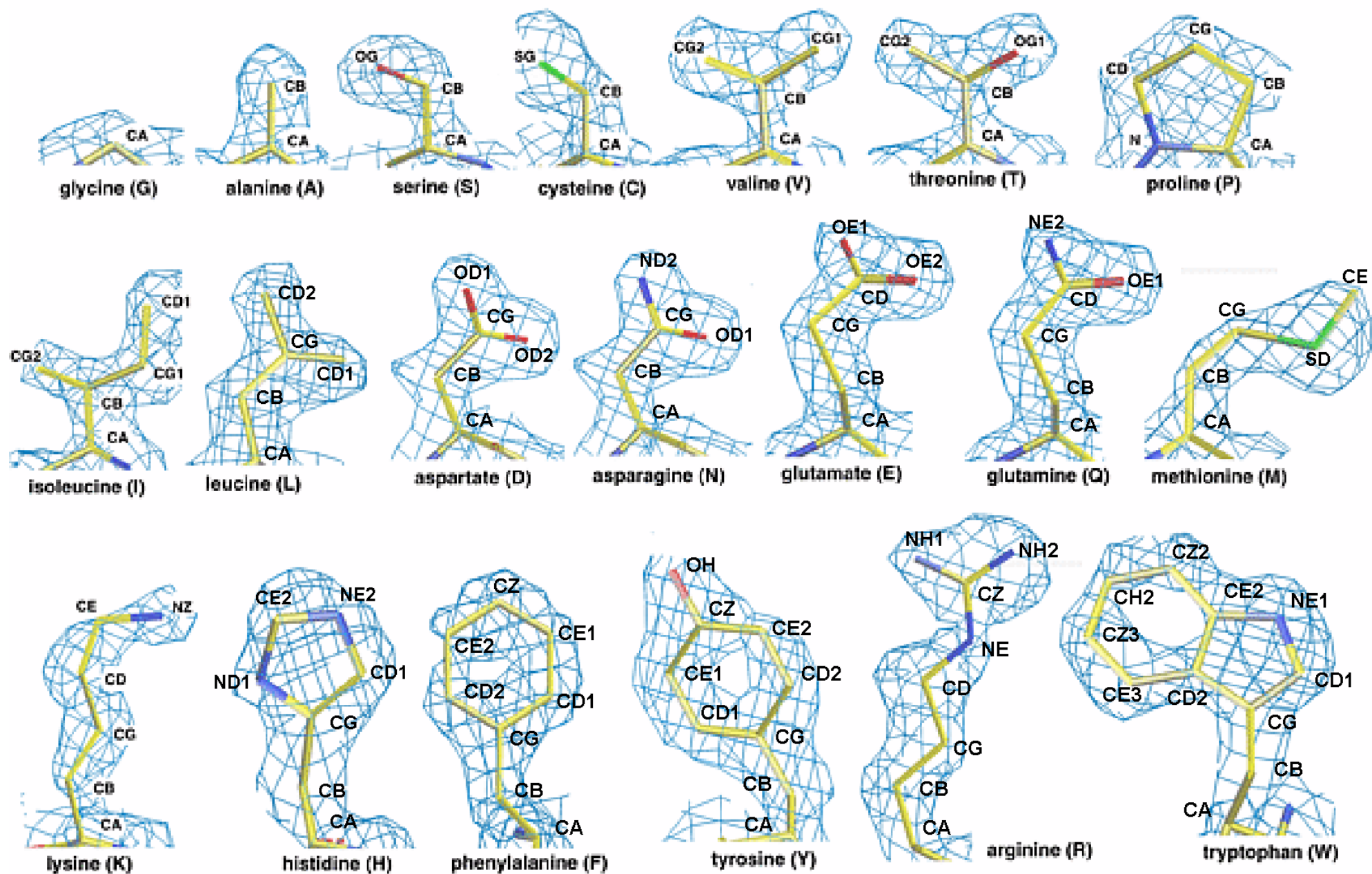
Protein structures

- Elements of structure
 - The amino acids
 - Levels
 - Databases
 - Folds and families
- Prediction
- How to check for correct assignment
 - Ramachandran plot

Structures of biomolecules

- Primary structure
 - Amino acid sequence
- Secondary structure
 - Local elements
 - Helices
 - Sheets
- Tertiary structure (3D)
 - Fold
 - Classification
- Quarternary structure
 - Interactions between chains
 - Protein-protein interaction

AMINO ACIDS

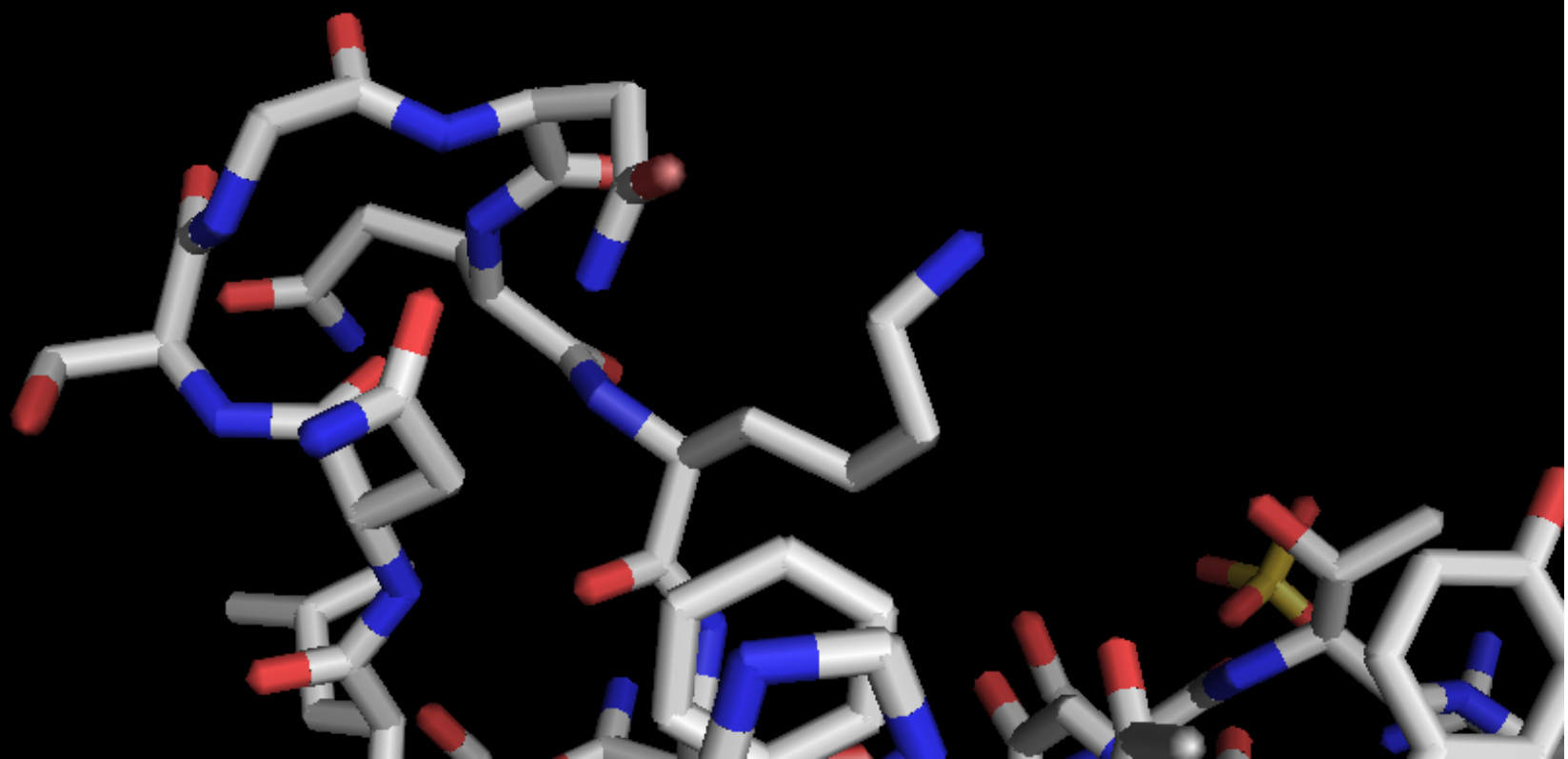


Visualizing Proteins

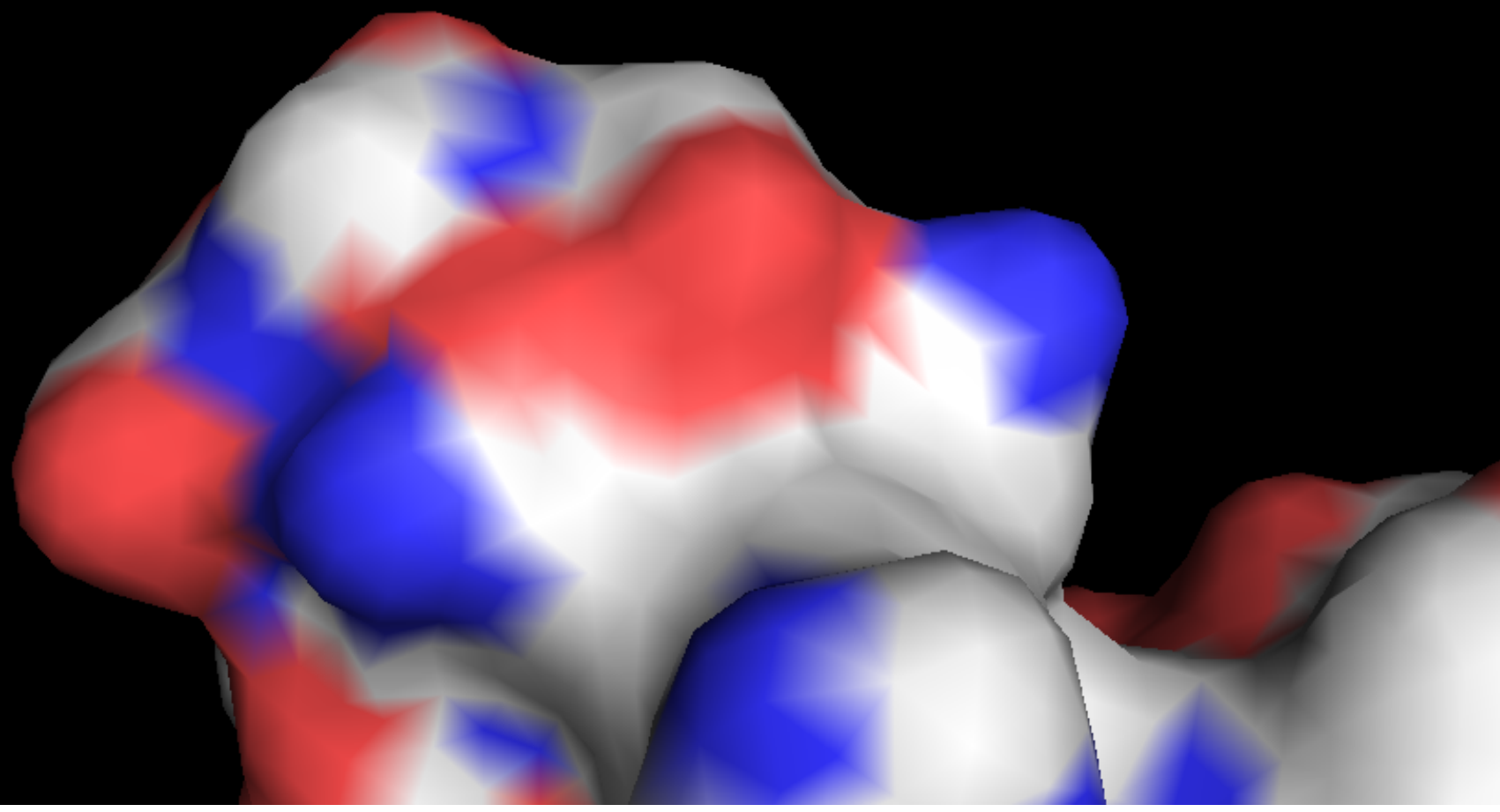
- High complexity
- Multiple levels of structure
- Important properties are “distributed
- throughout the 3D structure

- No single/simple “point” at which to look

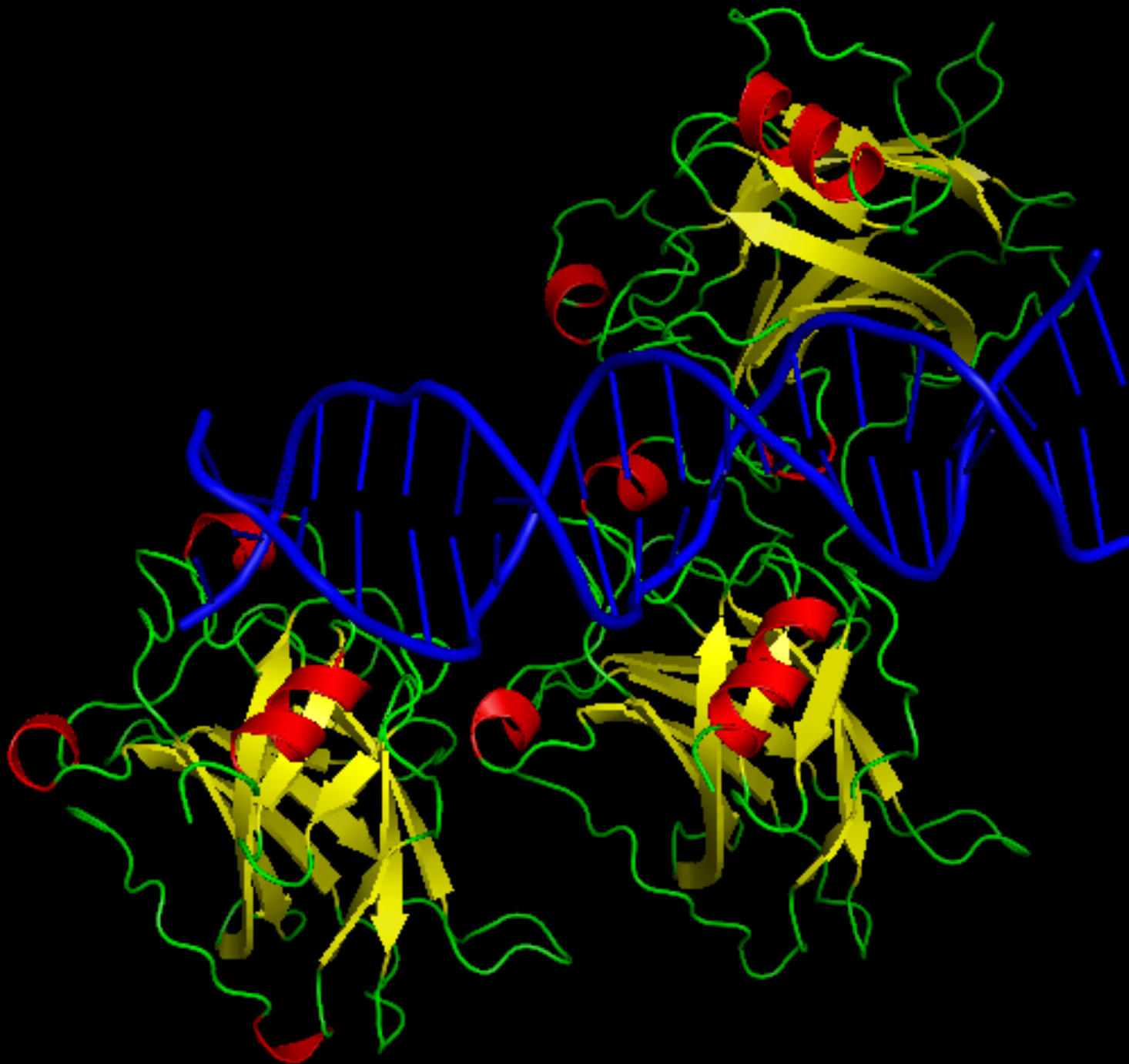
Wireframe

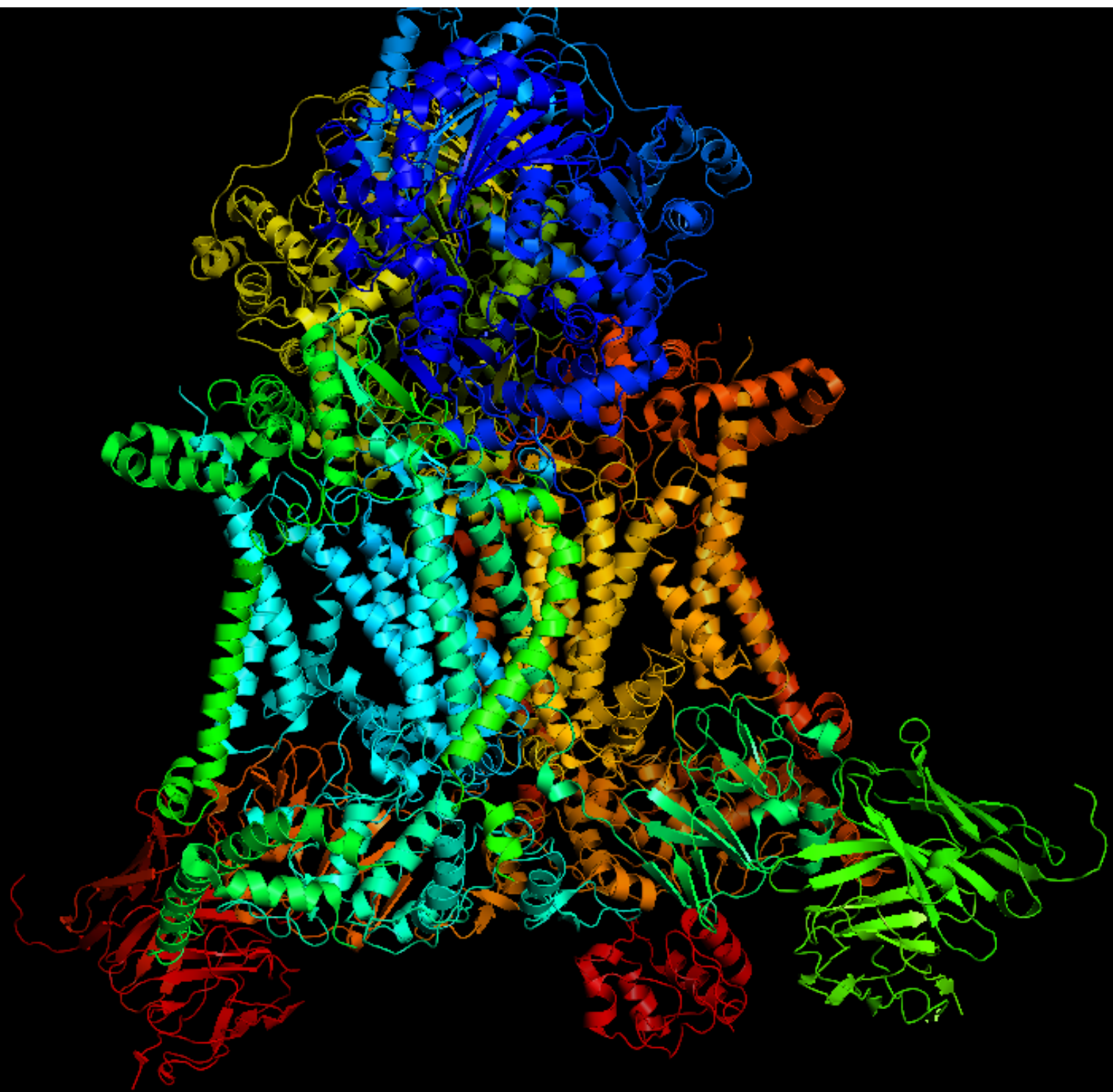


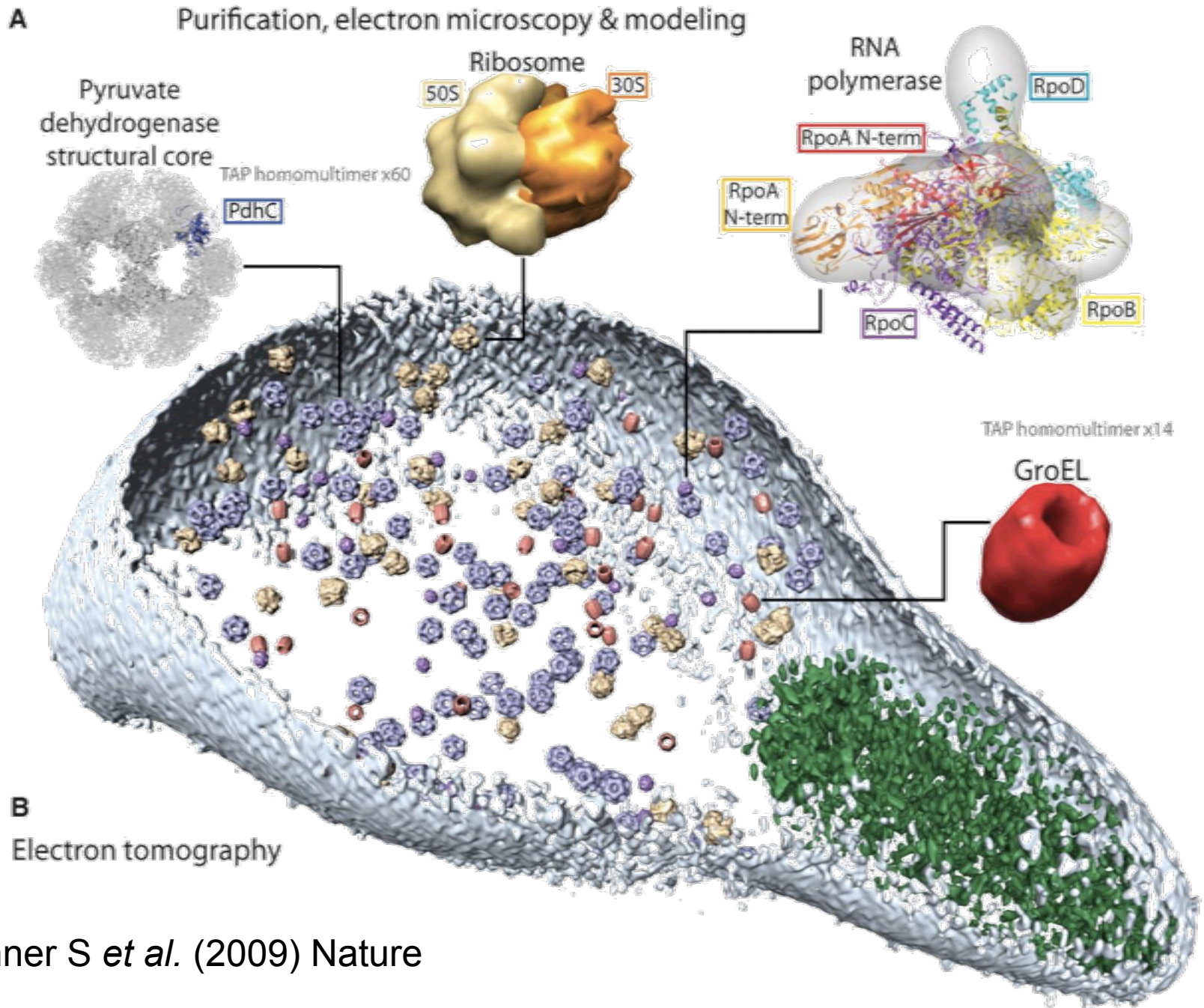
Surface



Cartoon







Kühner S *et al.* (2009) Nature

PDB/RCSB database

- Protein Data Bank – One of the oldest databases on molecular biology
- Repository of all known structures
 - All published structures must be deposited
- Four-character identifier

Classification of protein structures

CATH database

- Fold
- Superfamily – Secondary structure contacts
- Sequence families
- **Domains**

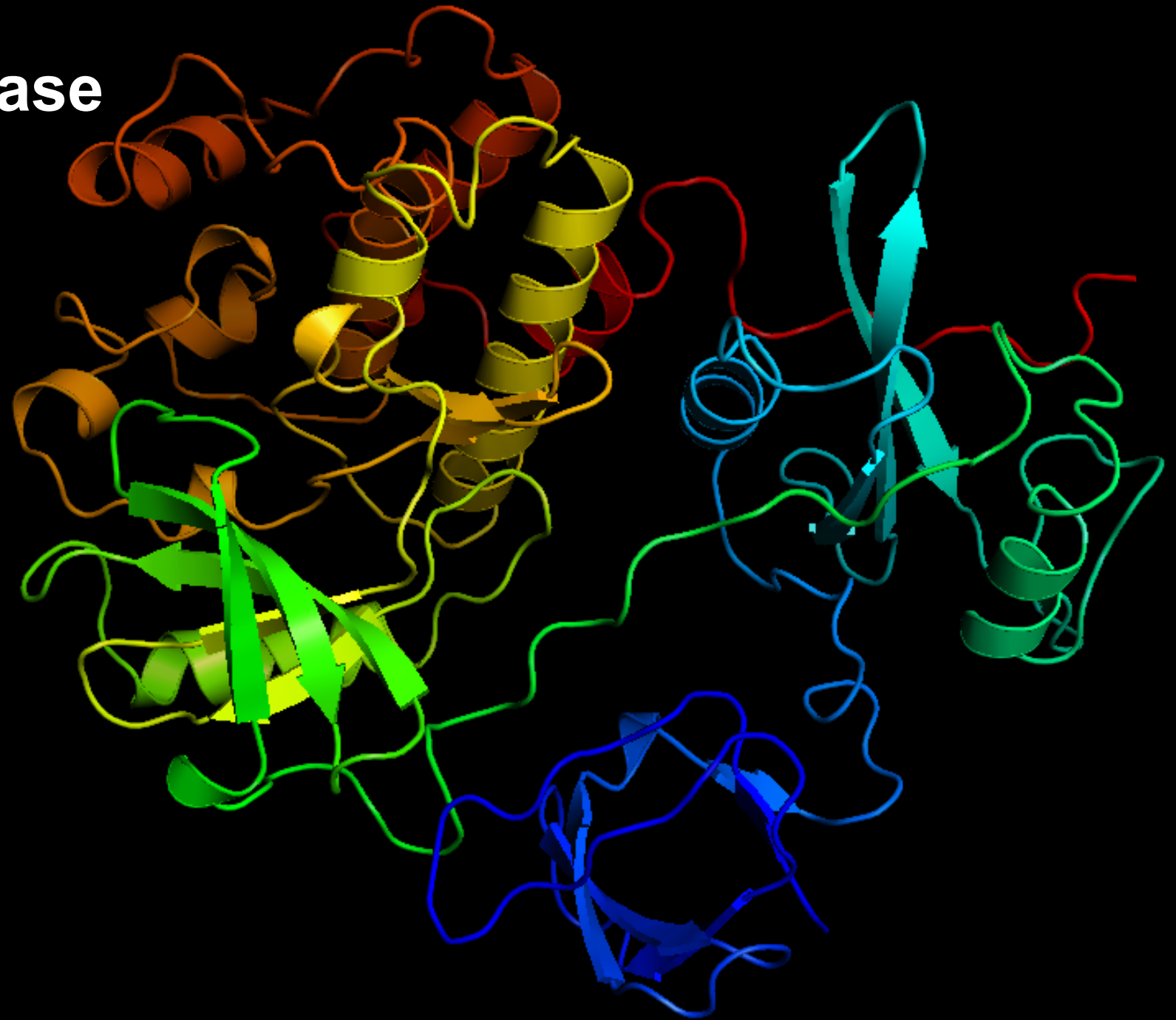
- Rule based on secondary structure content, contacts and domain boundaries

SCOP database

- Class
 - All α , all β
 - α / β – Parallel sheets
 - $\alpha + \beta$ – Antiparallel sheets
 - **Multi-domain proteins**
 - Membrane
 - Unstructured proteins

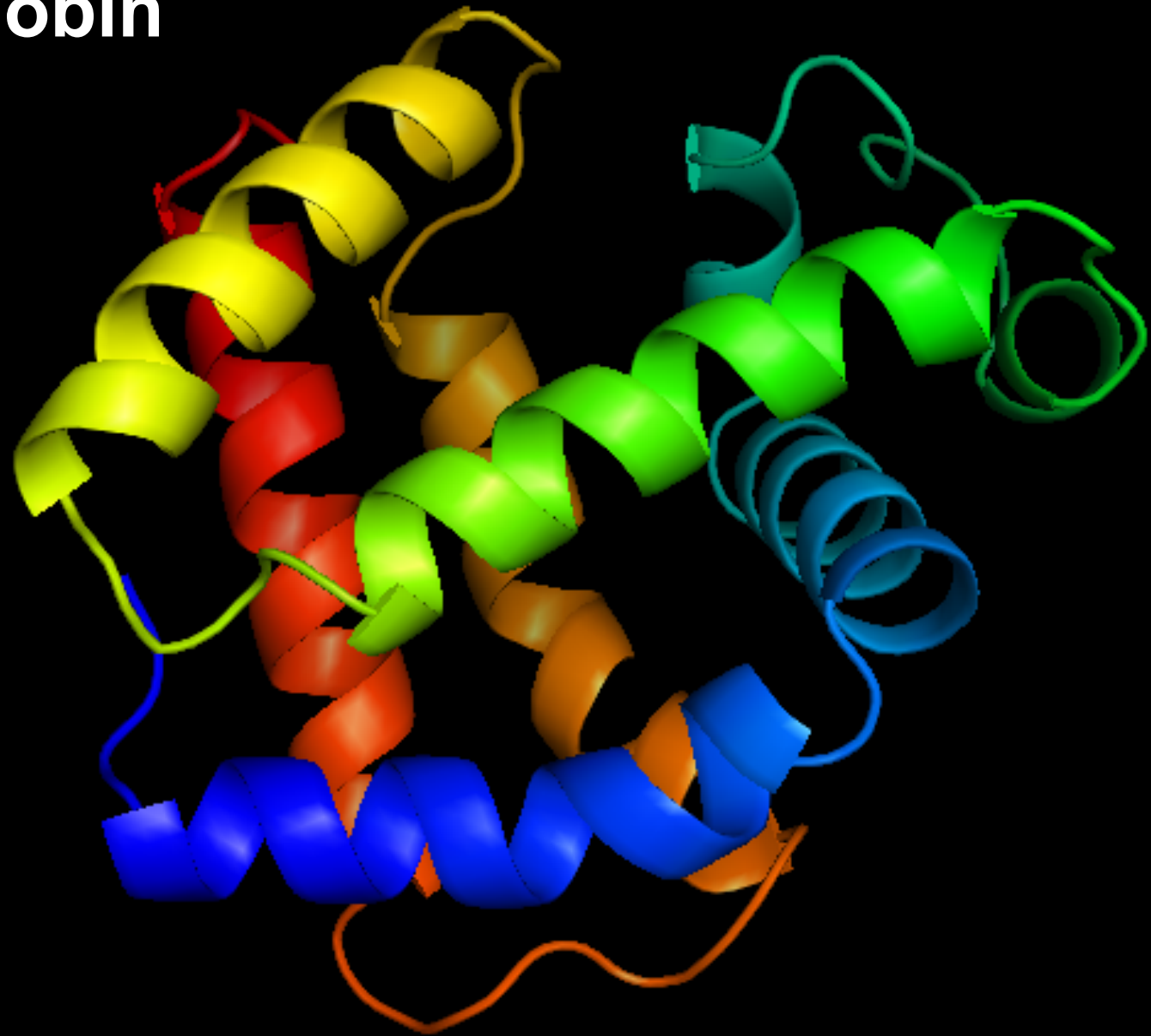
- Folds
- Superfamilies
- Families

SRC kinase



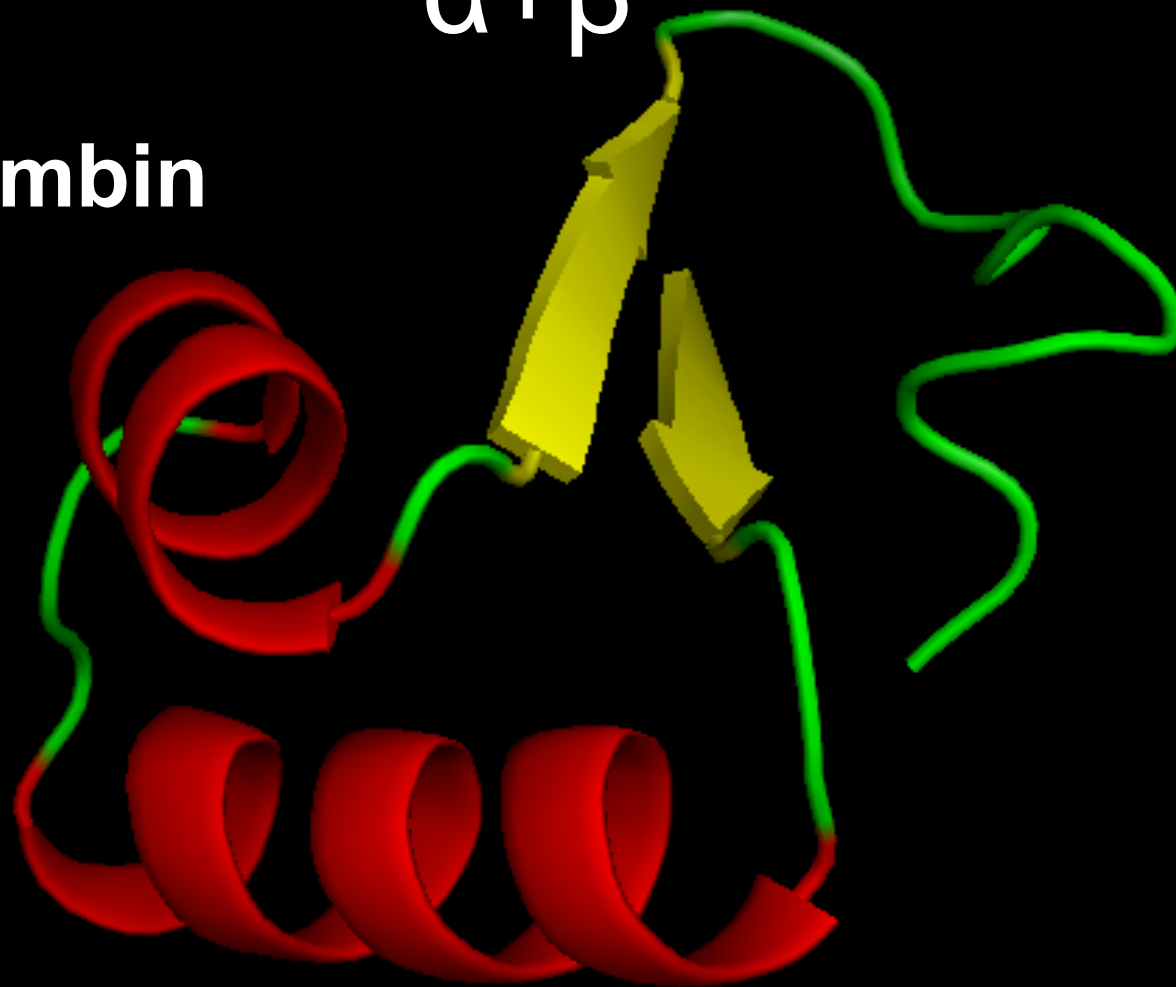
FOLDS

Hemoglobin

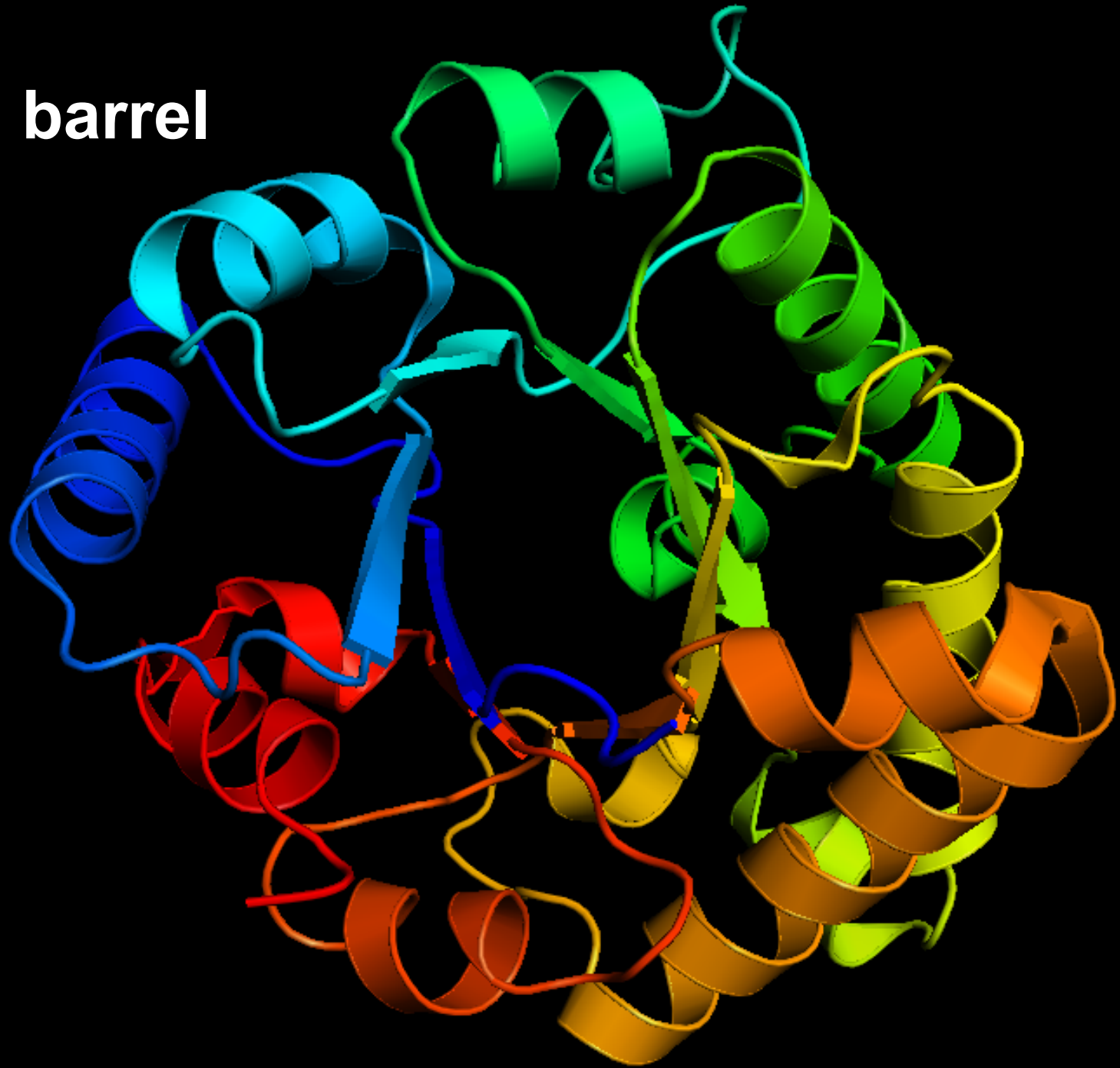


Crambin

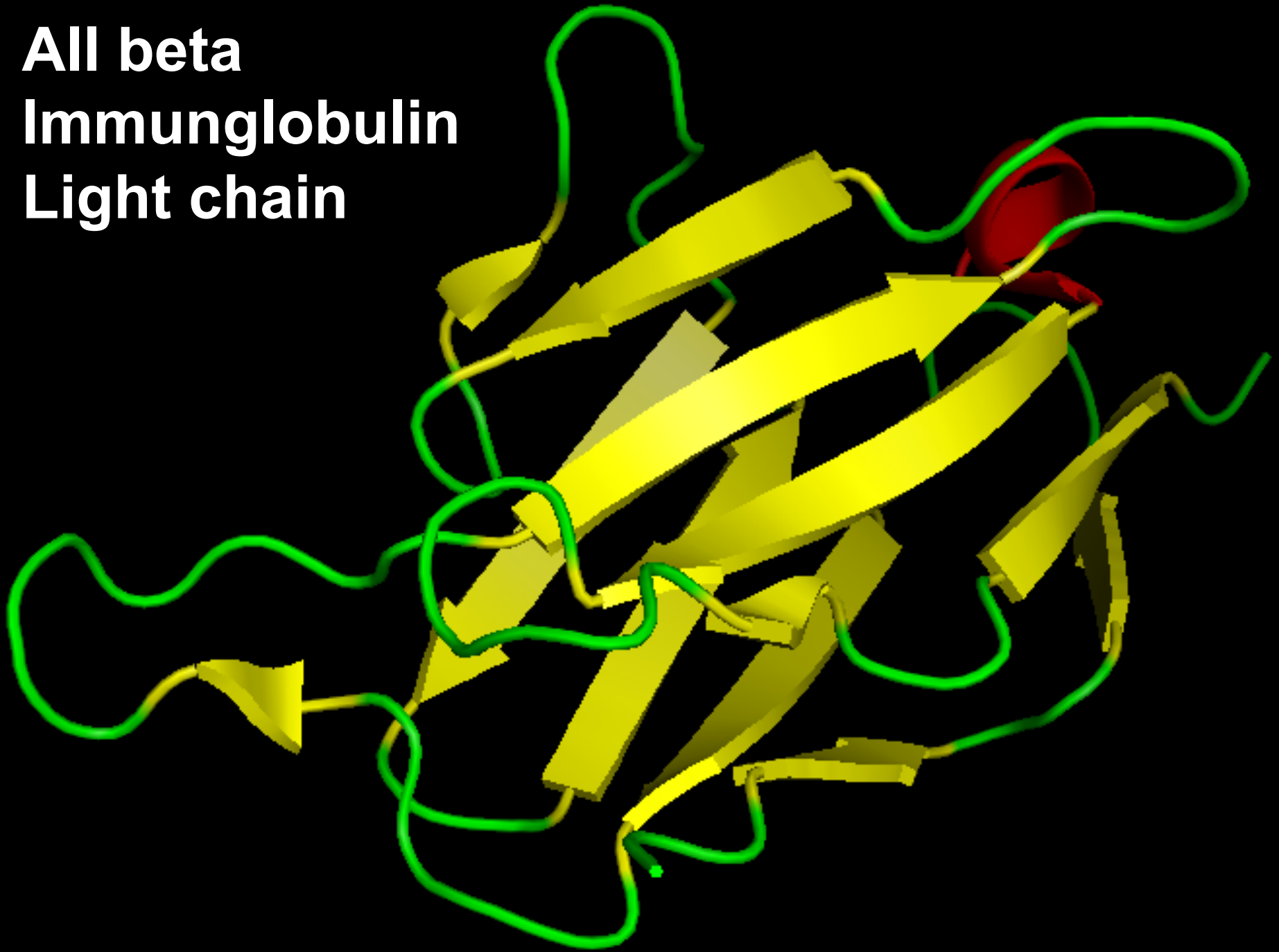
$\alpha+\beta$



α/β TIM barrel



**All beta
Immunoglobulin
Light chain**



PREDICTION OF PROTEIN STRUCTURES

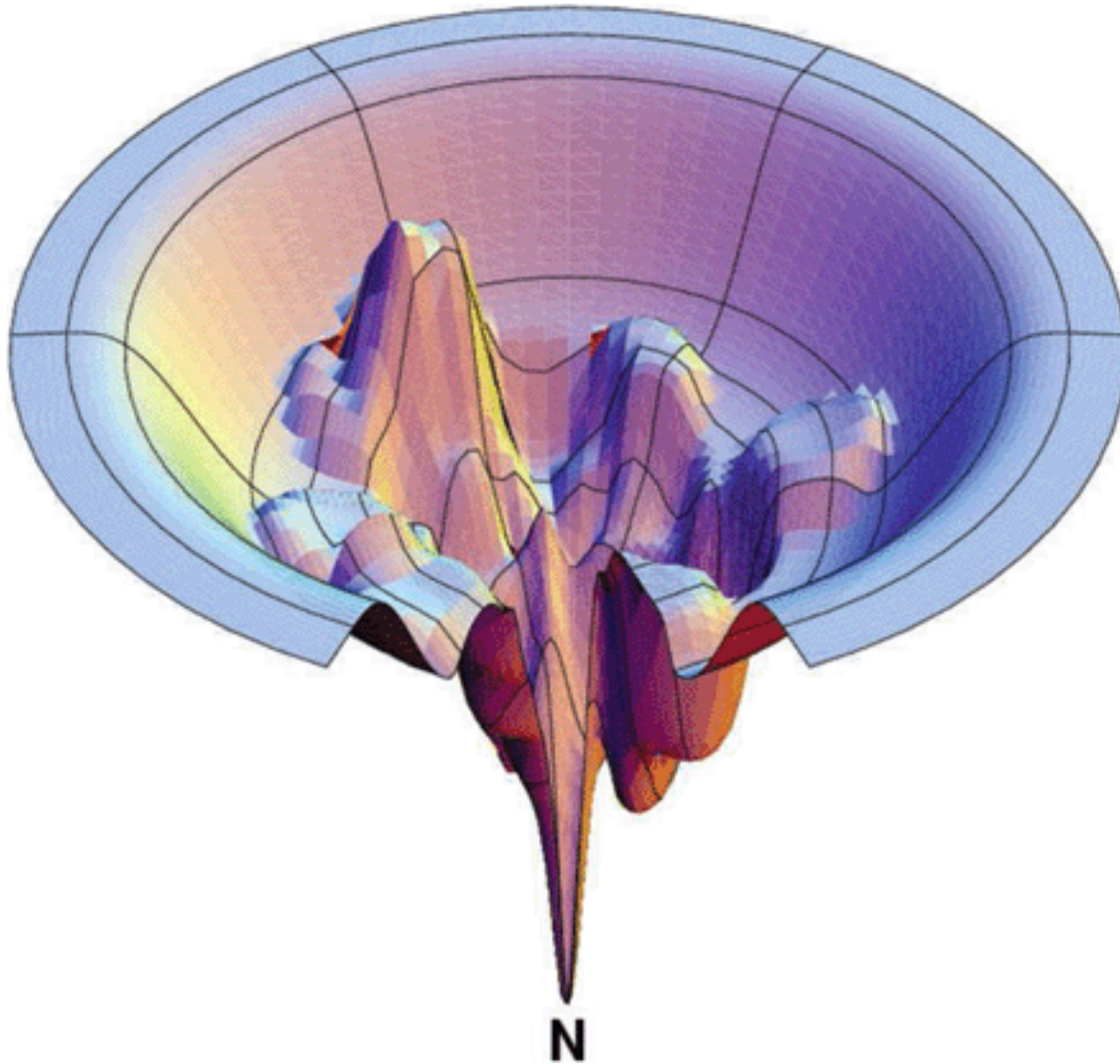
Anfinsen's dogma (1961)

- Denatured proteins can refold *in vitro*
- No folding machinery required
- All information about the structure resides in the sequence
- Native structure: minimum free energy
 - Unique
 - Stable
 - Kinetically accessible

Levinthal's Paradox

- Consider a protein with 101 residues
 - 100 Ψ and 100 ϕ angles
 - If we assume only three stable positions and none for ω
 - 3^{200} or 10^{95} conformations
 - Sampling all conformations exceeds the life time of the universe
- Proteins fold in milliseconds

Folding landscape



Secondary structure

- Single sequence methods
 - Chou-Fasman
 - GOR
- Neural networks
 - PHD
- HMMs

Chou-Fasman

Name etc	P(a)	P(b)	P(t)	f(i)	f(i+1)	f(i+2)	f(i+3)
Alanin	142	83	66	0.06	0.076	0.035	0.058
Threonie	83	119	96	0.086	0.108	0.065	0.065
...							

Calculate if $P(a) > 100$ for 4 out 6 AA, assign helix
Calculate if $P(b) > 100$ for 3 out 5 AA assign sheet
Calculate $p(t) = f(i) \dots$ assign turn
Further rules to resolve clashes

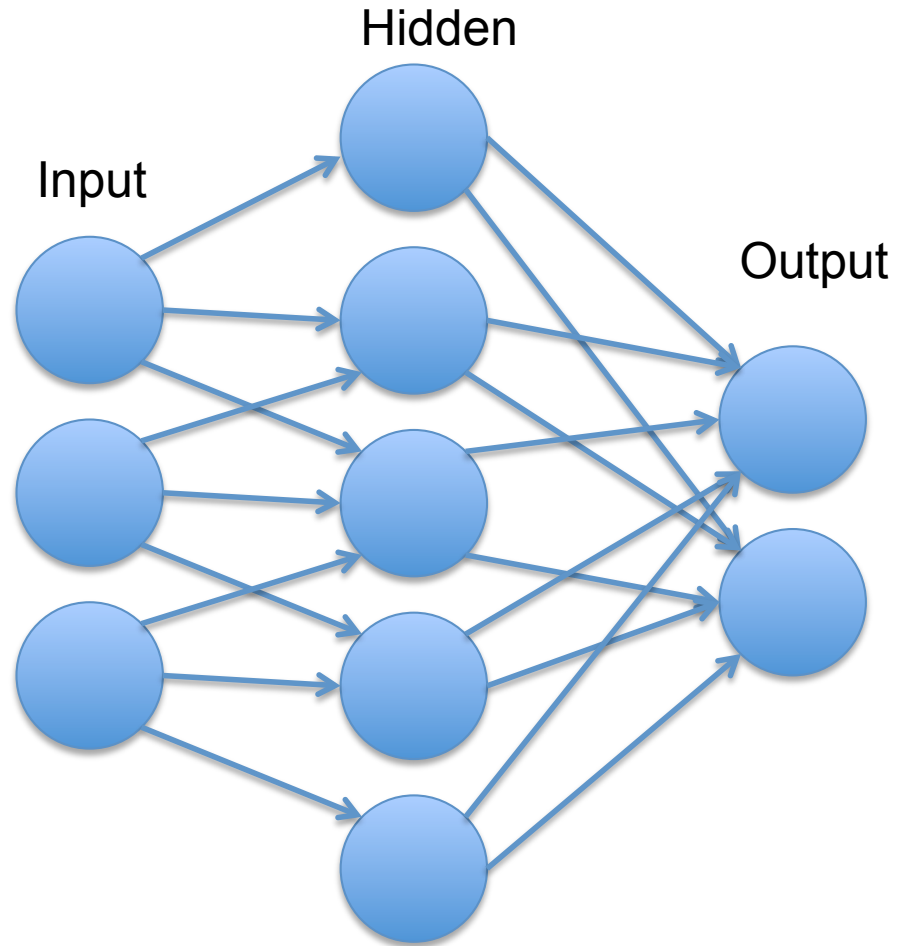
Chou and Fasman (1974) Biochemistry

Single sequence methods

- Prediction based on propensity of an AA to occur in helix, sheet or turn
- Chou-Fasman
 - Empirical, rule based
- GOR
 - Log-odds score, Bayesian statistics

Neural network

Machine learning technique
inspired by neuronal structures



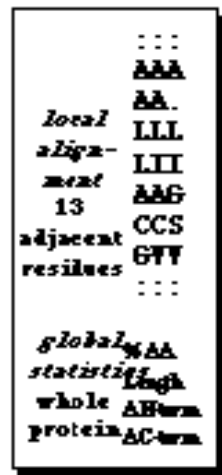
PHD

sequence information from protein family

profile derived from multiple alignment for a window of adjacent residues

two levels of neural network systems: PHDsec and PHDhtm

one level network: PHDacc

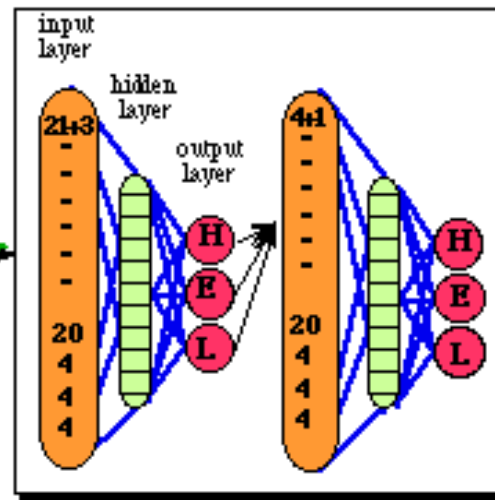


input local in sequence

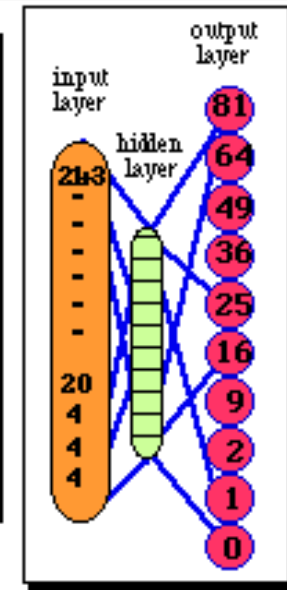
A	C	L	I	G	S	V	ins	del	cons
100	0	0	0	0	0	0	0	0	1.17
100	0	0	0	0	0	0	33	0	0.42
0	0	100	0	0	0	0	0	33	0.92
0	0	33	66	0	0	0	0	0	0.74
66	0	0	0	33	0	0	0	0	1.17
0	66	0	0	0	33	0	0	0	0.74
0	0	0	33	0	0	66	0	0	0.48

input global in sequence

percentage of each amino acid in protein
 length of protein (≤60, ≤120, ≤240, >240)
 distance: centre, H-term (≤40, ≤30, ≤20, ≤10)
 distance: centre, C-term (≤40, ≤30, ≤20, ≤10)



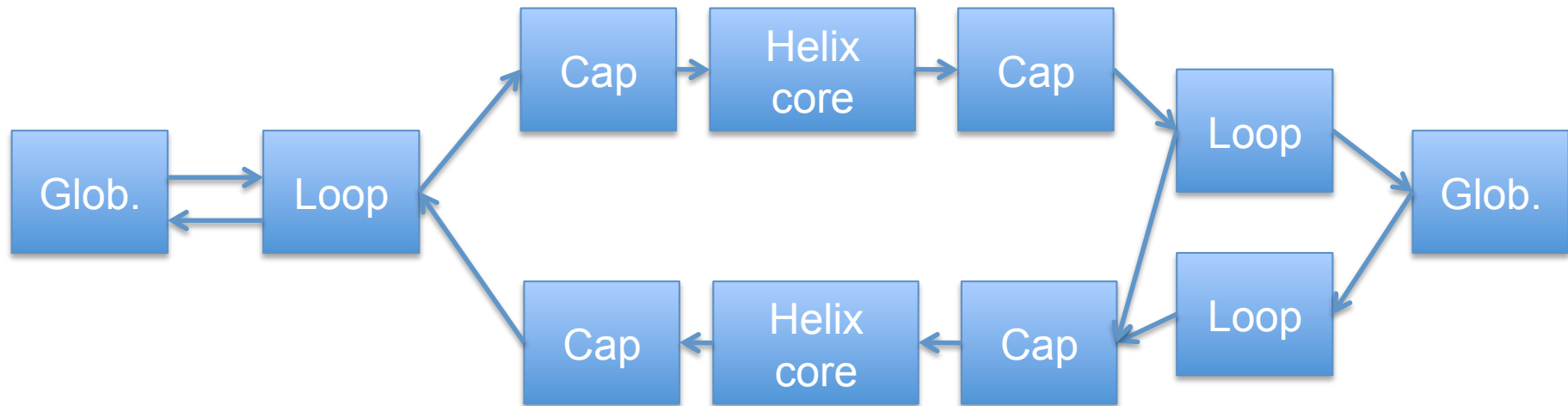
first level: sequence-to-structure network
 second level: structure-to-structure network



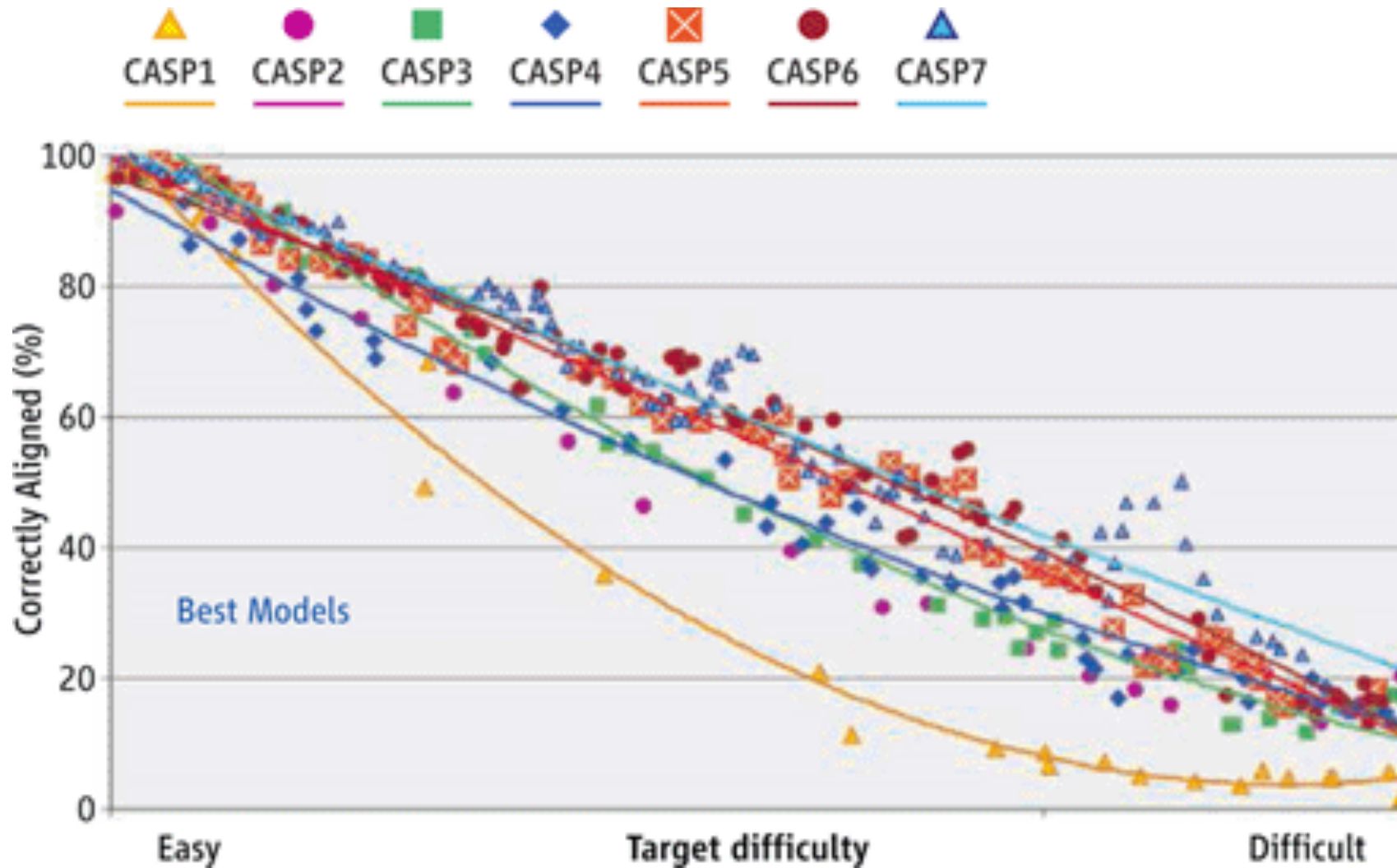
first level only



TMHMM



CASP



A. KRYSHTAFOVYCH ET AL., PROTEINS: STRUCTURE, FUNCTION, AND BIOINFORMATICS (5 OCTOBER 2007)

Tertiary structure

- Homology modeling
- Threading
 - Fold recognition
- Ab initio modeling

RMSD

- Root-mean-square deviation
- Distance of backbone atoms
 - Usually α

$$RMSD = \sqrt{\frac{1}{N} \sum_{i=1}^N \delta_i^2}$$

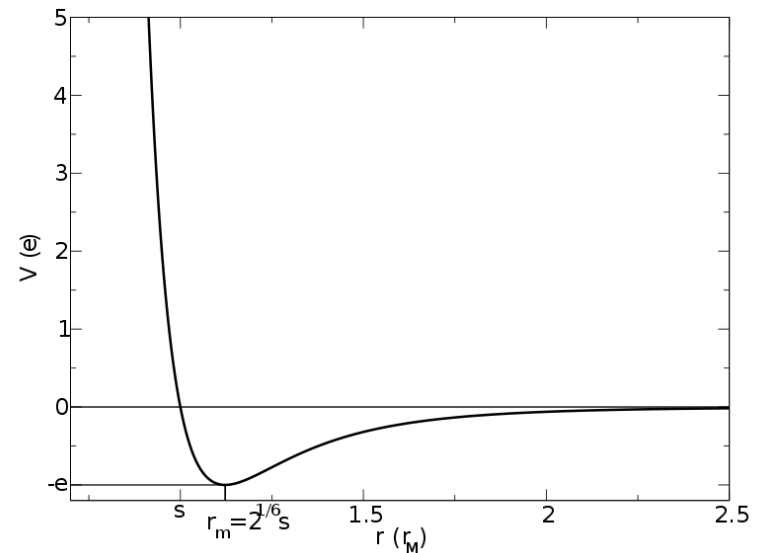
Some chemistry

- Intramolecular forces
 - Covalent bonds (400 kcal)
 - Strong but only relevant for cystin
- Intermolecular forces
 - Hydrogen bonds (12 – 16 kcal)
 - Van der Waals forces
 - Dipole-dipole (0.5 -2 kcal)
 - London (<1 kcal)
 - Buried hydrophobic faces

Lennard-Jones potential

- Summarizes the repulsion of atoms and attraction by van der Waals forces

$$V_{LJ} = 4\epsilon \left[\left(\frac{\sigma}{r} \right)^{12} - \left(\frac{\sigma}{r} \right)^6 \right]$$
$$= \epsilon \left[\left(\frac{r_m}{r} \right)^{12} - 2 \left(\frac{r_m}{r} \right)^6 \right]$$



Source: Wikipedia. Lennard-Jones potential

Structure prediction

1. Find backbone structure
 1. Homology modeling
 2. Threading
 3. Ab initio prediction
2. Loop modeling
3. Sidechain packing
4. Refinement

Homology modeling

- Find homologous sequence (BLAST etc)
- Multiple alignment (Muscle etc)
- Replace backbone in defined, conserved parts
- Check core model and re-align
- Model side chain
- Model loop regions
- Energy minimization

Homology modeling

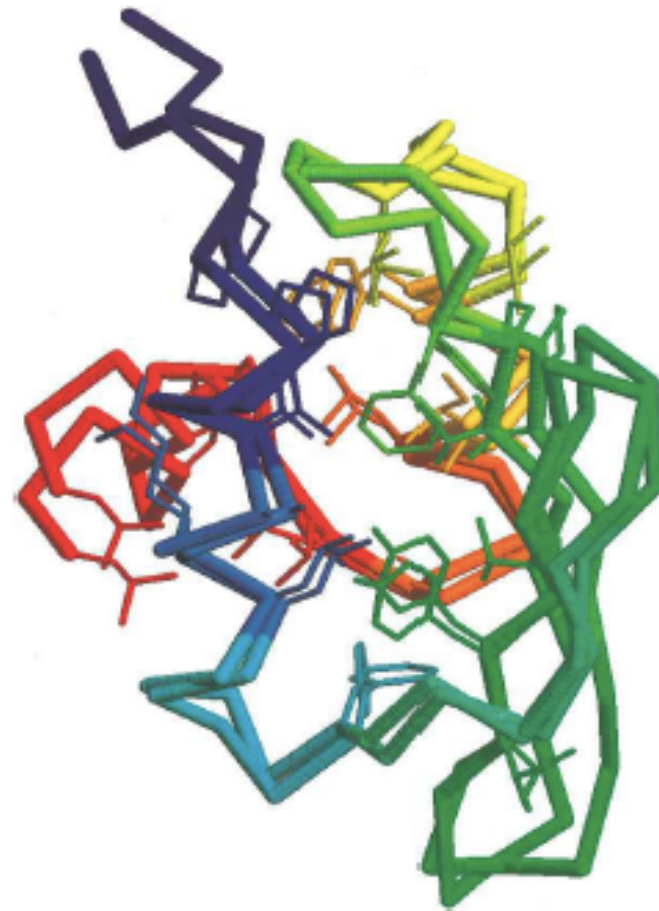
- Simple procedure for ID>40% over 50 AA (typical values, check for plausibility)
- Difficult if ID <25% over reasonable range
- Automated, SWISSMODEL available for all suitable targets
- If no template can be found:
 - Search template with sensitive methods: threading
 - Build from scratch: *ab initio*

Threading

- Naïve approach: Perform Homology Modeling for many/all templates, score the best
- Alignments at low %ID become problematic
- Fold recognition occasionally works, models often fail

Ab initio prediction

- Library of k-mers from known structures
- Build „random“ structures of k-mers
- Optimize in cycles, using a custom scoring function
- Analyze the top structures according to protein-like appearance and/or expectations from the literature.
- ROSETTA (Baker et al. (1998) outperformed contestants in CASP3.



Problem solved?

- Great improvements for globular proteins
- Open issues
 - Membrane proteins
 - Unstructured regions
 - Large assemblies