

Protein structure alignments

Proteins that fold in the same way, i.e. "have the same fold" are often homologs.

Structure evolves slower than sequence

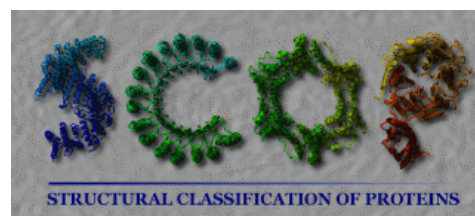
Sequence is less conserved than structure

If BLAST gives no homologs (*i.e.* sequence based)

Instead: Search with protein *structure* (pdb-file) in *structure database* (e.g. PDB) to find more remote homologs

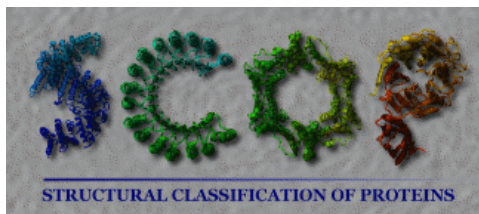
- For example using DALI
- Much more sensitive than sequence search
- Problems
 - Much smaller database (PDB vs. Genbank)
 - Need 3D structure of protein

Use structure comparisons to classify, group and cluster proteins. Build protein structure families and hierarchies



Protein structure classification

- Based on taking all structures of PDB
- Remove redundancy (*i.e.* keep only one copy of “identical” structures)
- Split structures into domains
- Group domains/proteins based on similarity
- Two main classification schemes: SCOP & CATH



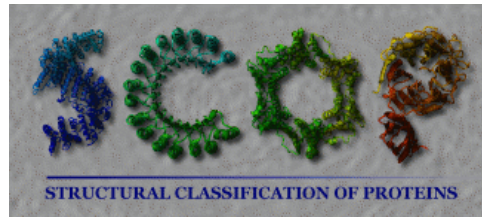
Structural Classification of Proteins

- Almost 100% manually generated
- Proteins grouped into hierarchy of classes, folds, superfamilies and families

Scop Classification Statistics

SCOP: Structural Classification of Proteins. 1.73 release
34494 PDB Entries (26 Sep 2007). 97178 Domains. 1 Literature Reference
(excluding nucleic acids and theoretical models)

Class	Number of folds	Number of superfamilies	Number of families
All alpha proteins	259	459	772
All beta proteins	165	331	679
Alpha and beta proteins (a/b)	141	232	736
Alpha and beta proteins (a+b)	334	488	897
Multi-domain proteins	53	53	74
Membrane and cell surface proteins	50	92	104
Small proteins	85	122	202
Total	1086	1777	3464



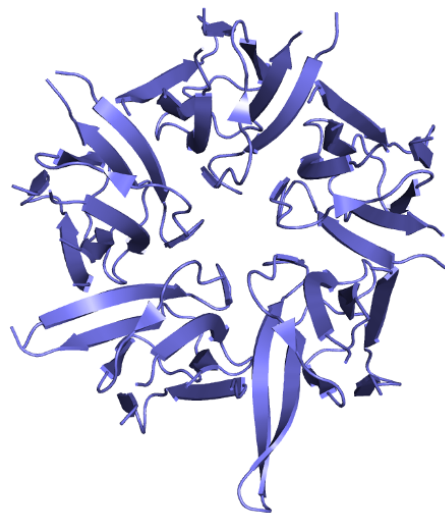
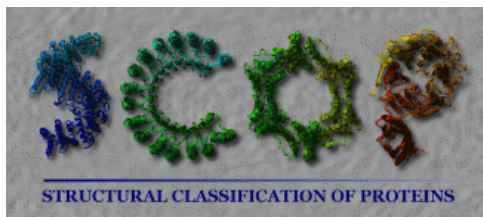
SCOP

Jon K. L  rdahl,
Structural Bioinformatics

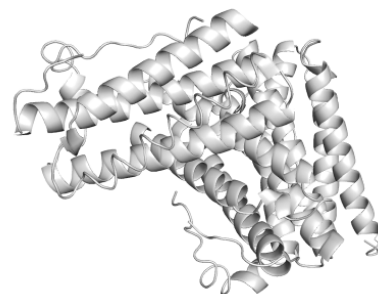
- Families
 - Sequence identity ~30% or higher
 - Very similar structures
 - Clearly homologous proteins
- Superfamilies
 - Contains families
 - May have no or little sequence similarity
 - Common fold
 - Are probably evolutionary related
- Folds
 - Contains superfamilies
 - Difficult level of classification
 - Same major secondary structure elements (α -helices and β -sheets) with same connections
 - Not always homologs
- Classes
 - Upper level of classification (4 major, 3 minor)
 - Contains folds
 - Based on secondary structure composition and “general features”
 - e.g. all- α , all- β , “membrane and cell surface” and “small proteins”
 - α/β : One β -sheet with strands connected by single α -helices
 - $\alpha+\beta$: α -helical and β -sheet part separated in sequence

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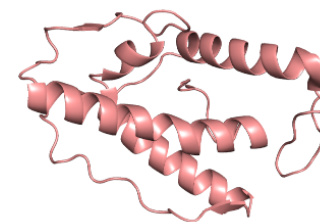
SCOP



all- β class



4-helical cytokines



T4 endonuclease V



Globin-like

all- α class,
3 different folds



TIM-barrel fold
 α/β class



Profilin-like fold
 $\alpha+\beta$ class



Class, Architecture, Topology
and Homologous

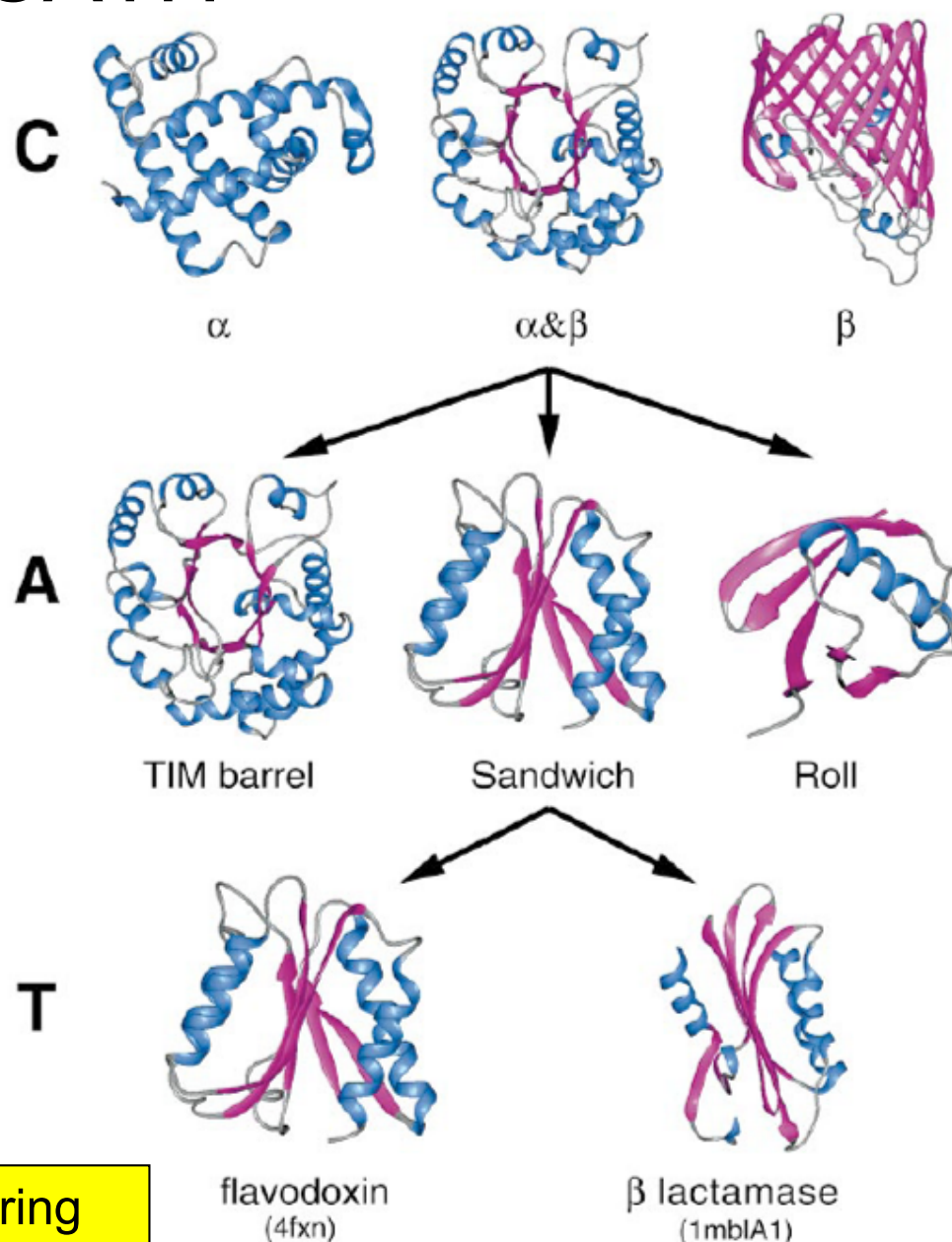
Both manual structural
alignment and automatic
alignment with SSAP

5 levels in hierarchy

- Class (as in SCOP)
- Architecture (unique to CATH)
- Fold/Topology (as in SCOP fold)
- Homologous Superfamily (as in SCOP)
- Homologous family (as in SCOP)

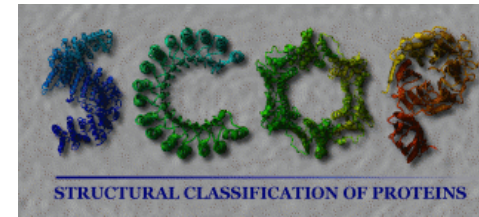
Explore during
the exercises??

CATH





CATH vs. SCOP

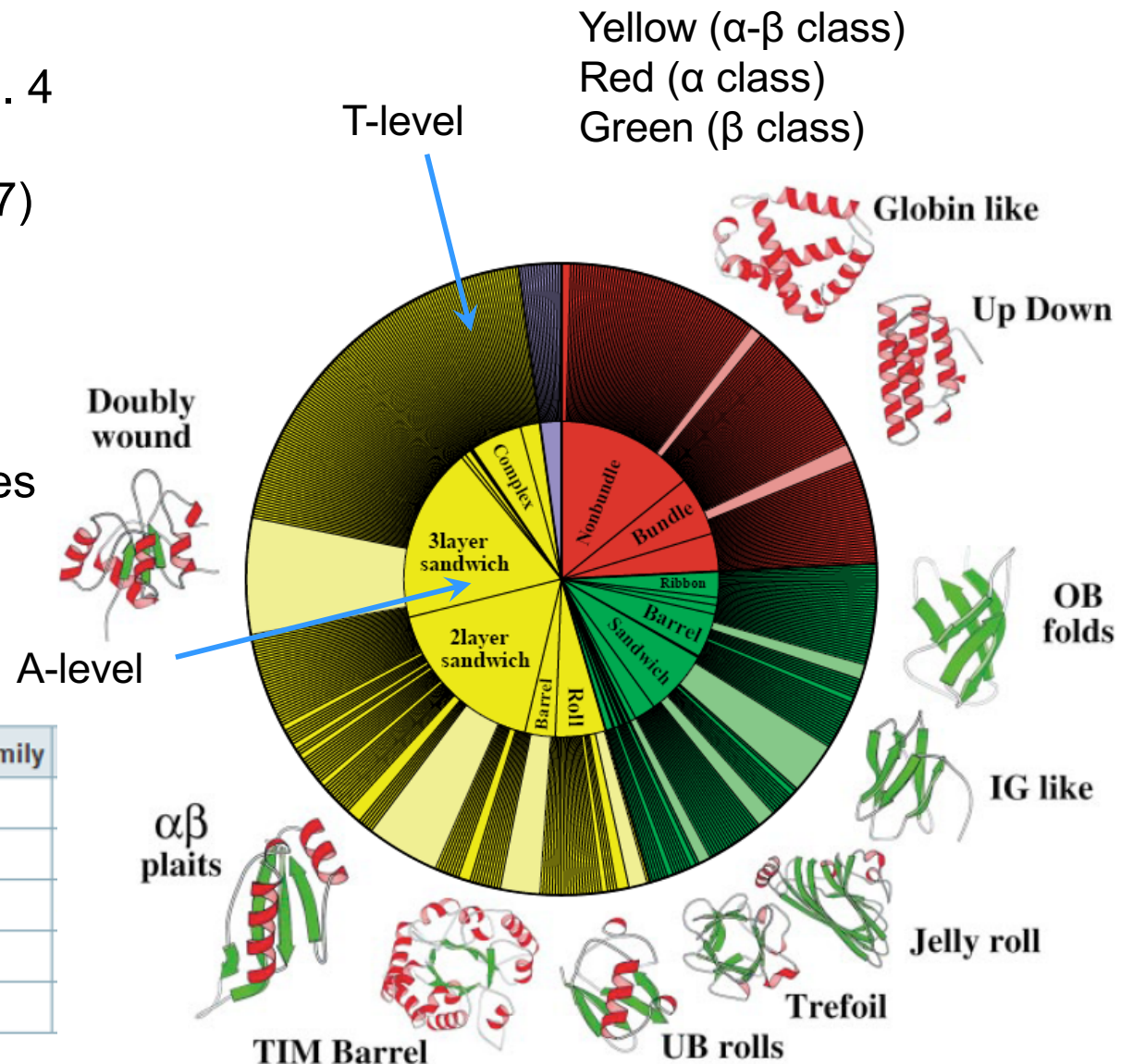


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Structural Bioinformatics

- Not always same domains
- Differences in hierarchy (5 vs. 4 levels)
- Differences in classes (4 vs. 7)
- Fully manual (SCOP) vs. manual/automatic (CATH)
- Most of the time (~80% of cases) classification is similar
- Both systems has weaknesses and strengths
- Use both!

CATH Version 3.2

Class	Architecture	Topology	Homologous Superfamily
1	5	310	682
2	20	196	438
3	14	512	956
4	1	92	102
Total	40	1110	2178



New topologies/folds are not found often!

C.A. Orengo *et al. Structure* 5, 1093 (1997)

SCOP2 & SCOPe

The screenshot shows the SCOP2 website with a blue header. The main content area includes a 'Welcome to SCOP2!' section, a 'Search Browser' with a search bar, and a 'Search Graph' section. A 'News' sidebar on the left lists updates from November 2013 to January 2014. The footer contains copyright information for 2014 MRC Laboratory of Molecular Biology.

Use this, most likely



The screenshot shows the SCOPe website with a purple header. It includes a 'Welcome to SCOPe!' section, a 'Browse' tab, and a 'Stats & History' tab. A large graph on the right shows the growth of SCOPe over time, with annotations for key milestones. The footer contains copyright information for 1994-2016 The SCOP and SCOPe authors.

Classes in SCOPe 2.06:

1. **a: All alpha proteins** [46456] (289 folds)
2. **b: All beta proteins** [48724] (177 folds)
3. **c: Alpha and beta proteins (a/b)** [51349] (148 folds)
4. **d: Alpha and beta proteins (a-b)** [53931] (385 folds)
5. **e: Multi-domain proteins (alpha and beta)** [56572] (69 folds)
6. **f: Membrane and cell surface proteins and peptides** [56835] (59 folds)
7. **g: Small proteins** [56992] (94 folds)
8. **h: Coiled coil proteins** [57942] (7 folds)
9. **i: Low resolution protein structures** [58117] (25 folds)
10. **j: Peptides** [58231] (133 folds)
11. **k: Designed proteins** [58788] (44 folds)
12. **l: Artifacts** [310555] (1 fold)

<http://scop.berkeley.edu>

Predictors

Prediction tools

- Predictors are available
 - on the web (in public web servers)
 - as (usually) free or commercial software
 - packaged in large (often commercial) software suites
- Predictors have been made for determining all kinds of features from sequence
 - Secondary structure
 - Structural disorder
 - Domain boundaries
 - Membrane protein or not
 - Number of transmembrane α -helices
 - Metal ion binding sites
 - Post-translational modifications
 - Phosphorylation sites
 - Cleavage sites
 - And many more
- Subcellular localization
 - Nuclear protein?
 - Secreted protein?
- Interaction with other proteins, DNA etc. (usually with some knowledge of 3D structure)

These tools are
often extremely
useful to biologists!


Example here is *secondary structure prediction* but similar or related methods/algorithms are used in most predictors

Secondary structure prediction

Assigning secondary structure is *not trivial* and there is *no single consensus method* even when 3D structure is known

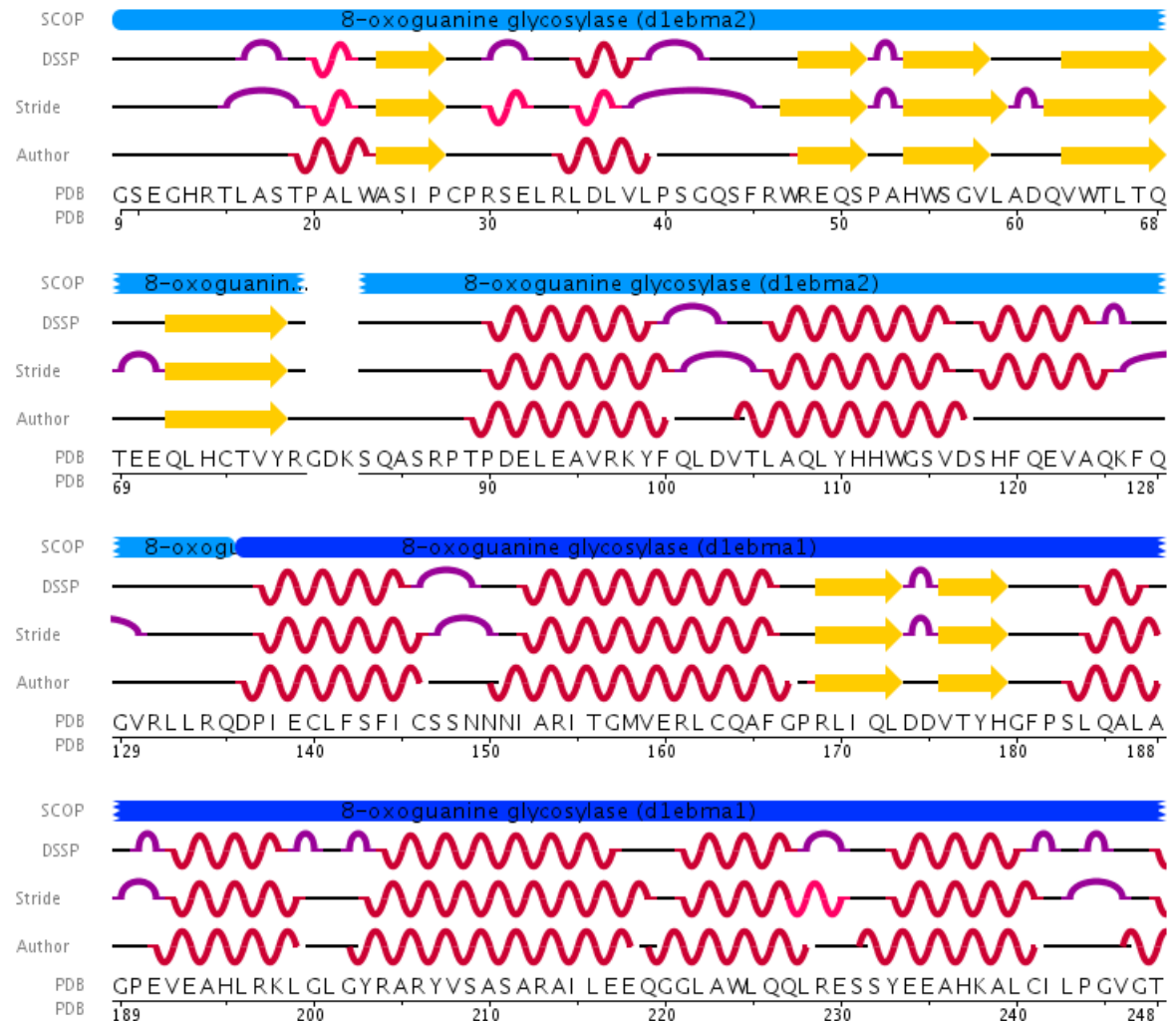
- Secondary structure may be put in manually by the authors behind a PDB-file
- Algorithms based on calculated H-bonds, Ramachandran plot, etc.

- DSSP
- STRIDE
- DEFINE

 β -strand
 α -helix

Everything else loop/coil

1EBM



Secondary structure prediction

Tools/programs that accept a primary sequence and predicts the secondary structure state (H/helix, E/sheet, or C/Loop&Coil) for each residue

The screenshot shows a web browser window titled "Department of Computer Science - Computational Biology Group: Prof - Windows Internet Explorer". The address bar shows the URL "http://www.aber.ac.uk/~phiwww/prof/". The browser's toolbar includes navigation buttons and a search bar containing "secondary structure prediction PROF". Below the toolbar, there are several tabs, with the active one being "Department of Computer Science - Computational Biol...". The main content area of the browser displays the website for the Aberystwyth University Computational Biology Group. The header features the university's logo and name, along with the group's name and address: "Aberystwyth University Computational Biology Group. Department of Computer Science, Aberystwyth SY23 3DB, Wales, UK". The main heading is "PROF - Secondary Structure Prediction System". Below this, there is a section titled "Submit a single amino acid sequence for secondary structure prediction:". This section contains a form with the following elements: a text input field for an email address, a red warning message "Please check twice, as we get a lot of predictions coming back, due to spelling mistakes!:", a dropdown menu for selecting the output format (currently set to "CASP"), a large text area for entering the amino acid sequence in FASTA format, and a "Submit Query" button. The browser's status bar at the bottom shows "Done" and "Internet" with a 100% zoom level.

Department of Computer Science - Computational Biology Group: Prof - Windows Internet Explorer

http://www.aber.ac.uk/~phiwww/prof/

secondary structure prediction PROF

Suggested Sites BioInfo Biology Journals Other Answers.com Bioinfo Links cbo-all Adm FUGE bioinf

Department of Computer Science - Computational Biol...

PRIFYSGOL ABERYSTWYTH UNIVERSITY

Aberystwyth University
Computational Biology Group.
Department of Computer Science, Aberystwyth SY23 3DB, Wales, UK.

PROF - Secondary Structure Prediction System

Submit a single amino acid sequence for secondary structure prediction:

Please specify your email address
Please check twice, as we get a lot of predictions coming back, due to spelling mistakes!:

Select your desired output format:
CASP

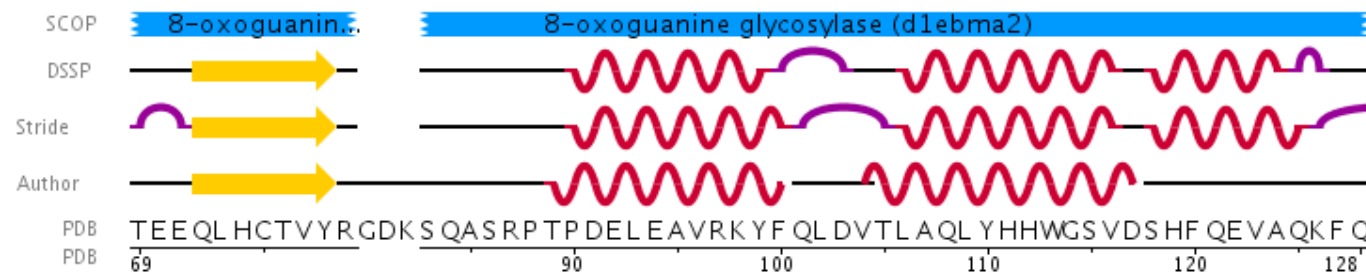
Please enter your sequence in FASTA format (first line starting with > and the title reference, followed by multiple lines of single letter amino acid sequence (NO ALIGNMENTS OR DNA PLEASE!!)):

Submit Query

Done Internet 100%

Secondary structure prediction

Tools/programs that accept a primary sequence and predicts the secondary structure state (H/helix, E/sheet, or C/Loop&Coil) for each residue



Human OGG1 TEEQLHCTVYRGDKSQASRP TPDELEAVRK YFQLDVTLAQLYHHWGSVDSHFQEVAQKFQ
 PROF Prediction CEEEEEEEEEC CCCCCCCCCCHHHHHHHHHHHH CCCCCCHHHHHHHHHHHH CCCCCCHHHHHHHHHHHH C

Uses:

- Correct and guide sequence alignments since secondary structure is more conserved than primary sequence
- Classify proteins
 - If you think your protein is a TIM-barrel, but your prediction suggests it has only α -helices, you probably are wrong
- Important step towards predicting 3D structure

Globular and transmembrane proteins have quite different properties and should be tackled with different algorithms

Secondary structure prediction

- Random prediction ~40% accuracy
- 1st generation prediction (1970's) ~50%
 - Based on relative *propensities*/intrinsic tendencies of each amino acid to be in a state X (= H, E, or C)
 - Ala, Glu & Met often in state H
 - Pro & Gly often in state C
- 2nd generation prediction (until mid 1990's) ~60%
 - Proper inclusion of propensities for neighboring residues
 - Larger experimental data set
- 3rd generation prediction (until present time) approaching ~80%
- Two main improvements:
 - Machine learning/neural networks
 - Combines information from predictions for single sequence with information from homologous sequences (multiple sequence alignment)

Since structure is more conserved than sequence homologs (>35% identity) are likely to have same secondary structure

Secondary structure prediction

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Structural Bioinformatics

- 3rd generation prediction (until present time) approaching ~80%
- Two main improvements:
 - Machine learning/neural networks
 - Combines information from predictions for single sequence with information from homologous sequences (For example sequences with >35% identity in multiple sequence alignment)

```
NP_833004/1-235 GNRKDNAFSESKISDMLEMKDTIHHSPT
1T06_Bc/1-256 GNRKDNAFSESKISDMLEMKDTIHHSPT
ZP_00740414/1-111 GNRKDNEFSESKISDMLEMKDTIHHSPT
ZP_00235456/1-229 GNRKDNEFSESKISDMLEMKDTIHHSPT
YP_052634/1-229 GNRKDNEFSESKISDMLEMKDTIHHSPT
ZP_00393536/1-235 GNRKDNEFSESKISDMLEMKDTIHHSPT
YP_037360/1-251 GNRKDNEFSESKISDMLEMKDTIHHSPT
NP_979598/1-235 GNRKDNEFSESKISDMLEMKDTIHHSPT
YP_084575/1-235 GNRKDNEFSESKISDMLEMKDTIHHSPT
YP_092361/1-235 GNRKDNEFSESKISDMLEMKDTIHHSPT
NP_712948/1-229 SILPNDQDSKEISKLLKRVESKTHKSNRV
YP_001221/1-235 SILPNDQDSKEISKLLKRVESKTHKSNRV
ZP_00533308/1-229 TKLHPERLNTKLTQSLLOKVEAQIPNAHNRV
ZP_00240774/1-227 -A1KNKTLQDDFFSPYLEEKWNTHNEKNRK
NP_832674/1-227 -A1KNKTLQDDFFSPYLEEKWNTHNEKNRK
NP_979281/1-227 -A1KNKTLQDDFFSPYLEEKWNTHNEKNRK
YP_037007/1-227 -A1KNKTLQDDFFSPYLEEKWNTHNEKNRK
ZP_00393174/1-227 -A1KNKTLQDDFFSPYLEEKWNTHNEKNRK
```

Predict
secondary
structure for all
these and fit
onto alignment

Generate prediction
based on consensus

Structure is more conserved
than sequence!
More sequences available
than structures (PDB vs
GenBank)!

Sequences
& *known* secondary structures
from PDB



Neural network is
trained on these
data

Sequences



Trained neural
network

Predicted secondary structures

Secondary structure prediction — consensus-based

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- Random prediction ~40% accuracy
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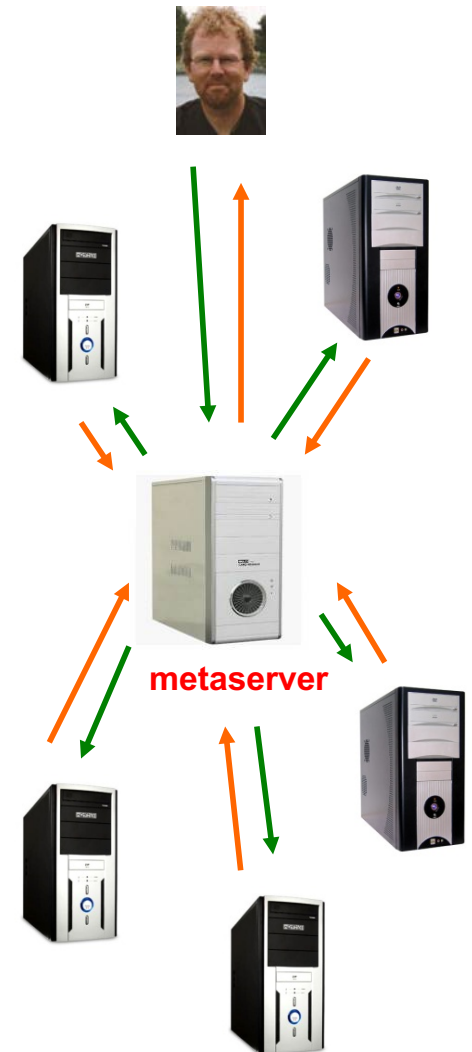
Many (more than 70 different published algorithms!) programs for secondary structure prediction:

- **PHD** – BLASTP to find homologs, MSA of homologs, neural networks used for prediction, web server
- **PSIPRED** – PSI-BLAST for homologs, MSA generated, neural network prediction, filtering, web server
- **PROF** – PSI-BLAST, MSA, neural network

Very good idea to use *not one tool* and trust the results, but instead use *several unrelated tools* and compare/use the consensus

Some web servers do this automatically and generates a consensus based on several algorithms (e.g. Jpred & PredictProtein)

- Several programs run and the results are presented to the user as
 - one consensus result
 - all results and the interpretation is left to the user
- The individual programs may be
 - run locally
 - on web servers other places on the internet with the results collected and combined on the consensus-server (**metaserver**)



→ Job query
→ Prediction result

Secondary structure prediction — consensus-based

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```

OrigSeq      : 1-----11-----21-----31-----41-----51-----61-----71-----81-----91 :
OrigSeq      : MSLPSLDSVPMLRRGFRFQFEP AQDCHVLLYPEGMVKLNDSAGEILKLVDGRRDVAAIVAALRERFPEVPGIDEDILAFLEVAHAQFWIELQ : OrigSeq

jalign       : -----H-----EEEE-----HHHHHHHHHH-----H-HHHHHHHHHHH-----HHHHHHHHHHHHHH----- : jalign
jfreq        : -----HHHHHHHH-----EEEE-----HHH-HHHHHHHHH-----HHHHHHHHHHHH-----HHHHHHHHHHHHHHHH----- : jfreq
jhmm         : -----EE-----EEEE-----E-HHHHHHHHHHH-----HHHHHHHHHHHH-----HHHHHHHHHHHHHH-----EEE- : jhmm
jnet         : -----HHHHH-----EEEE-----EEE-HHHHHHHHHHH-----HHHHHHHHHHHH-----HHHHHHHHHHHHHH-----EEE- : jnet
jpssm        : -----HHH-----HHH-----EEE-----EEE-HHHHHHHHHHH-----HHHHHHHHHHHH-----HHHHHHHHHHHHHH-----EE- : jpssm

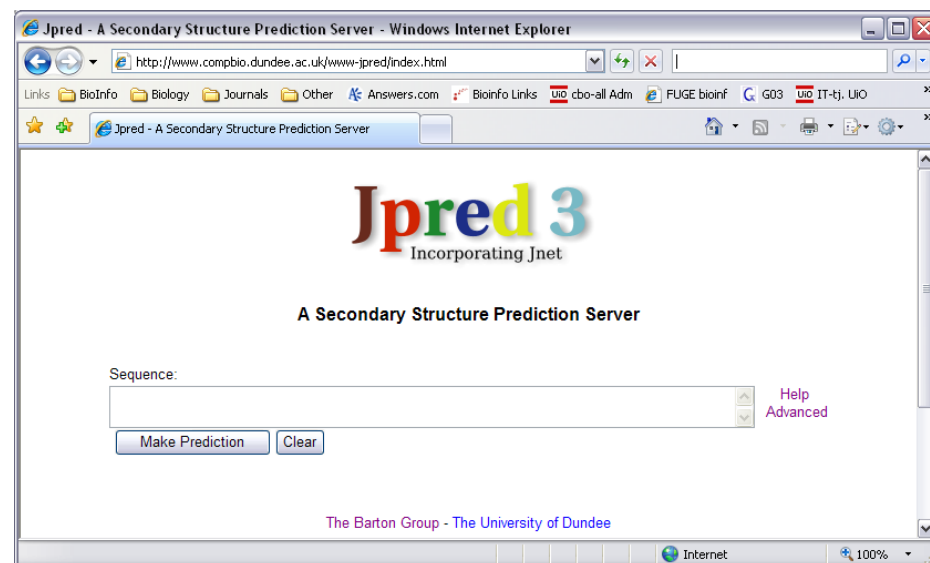
jpred        : -----HHHHH-----EEEE-----EE-HHHHHHHHHHH-----HHHHHHHHHHHH-----HHHHHHHHHHHHHH-----EEE- : jpred

Lupas 14     : ----- : Lupas 14
Lupas 21     : ----- : Lupas 21
Lupas 28     : ----- : Lupas 28

Jnet_25      : B--B---BBB-B-BBBB-BB-B-BBBBBBBB-BBBBBB-BBBBBB-BBBB-B-B-BB-B-B---B-BB-BB-B-B---BBB-B- : Jnet_25
Jnet_5       : -----B-B-B-----BBBBB-----B-B-B-BB-B-B---B-BB-B-B---B-BB-B-B---B-B- : Jnet_5
Jnet_0       : -----B-----B-----B-----B-----B-----B-----B-----B-----B-----B- : Jnet_0
Jnet_Rel     : 68888774110389831202254570799558841644325999998826841489999999997587998187899999998860525874 : Jnet_Rel
    
```

Puehringer *et al. BMC Biochemistry* 9:8 (2008)

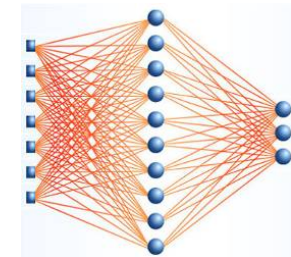
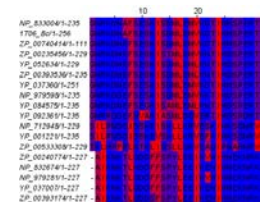
C. Cole *et al. Nucleic Acids Res.*
36, W197 (2008)



Predictors – *common features*

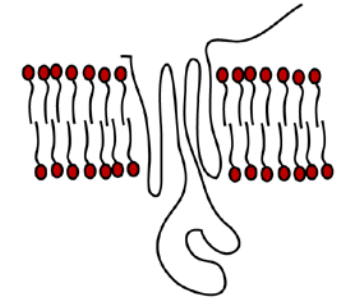
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- Use propensities/intrinsic tendencies of single residues or short sequence segments to be in a certain state (e.g. secondary structure state, order/disorder state, signal sequence)
- Include local interactions, *i.e.* take into account states in up- and downstream sequence
- Use homologous sequences to get predictions from many sequences with same structure/function
- Use neural networks or similar methods in predictions
- Consensus from many tools is better than just a single result (e.g. metaservers)

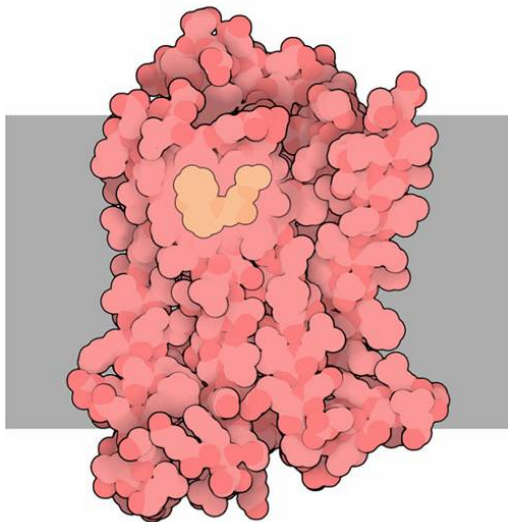


Transmembrane (TM) proteins

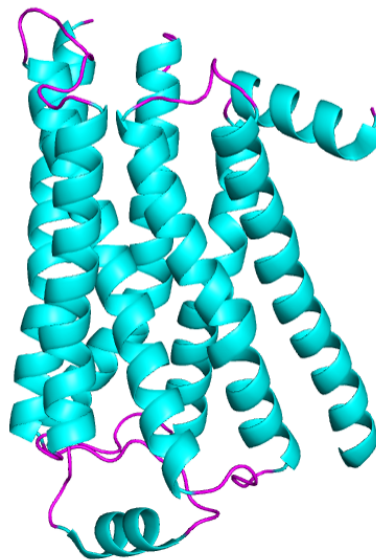
- ~30% of proteins in cells (but more than 50% of proteins interacts with membranes)
 - α -helical type: all membranes and organisms
 - β -barrel type: only outer membranes of Gram-negative bacteria, lipid-rich cell walls of a few Gram-positive bacteria, and outer membranes of mitochondria and chloroplasts



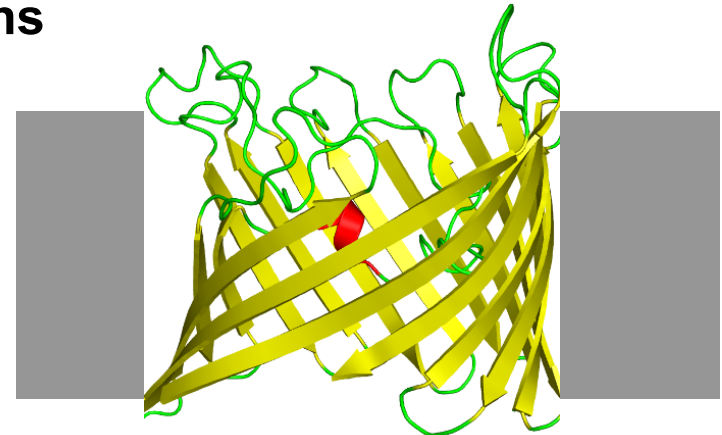
Can usually NOT use the same predictors for secondary structure and other properties as for globular proteins



PDB Apr. 08 "Molecule of the Month"



2RH1, Human adrenergic receptor



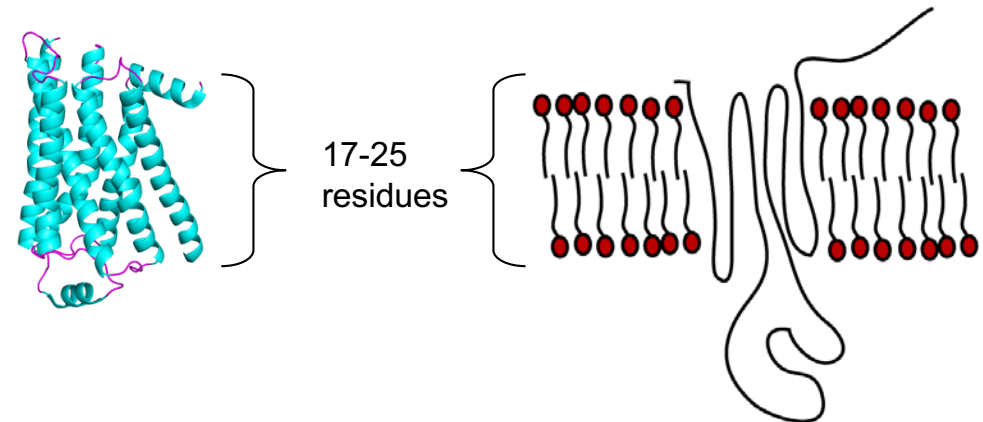
Porin

Transmembrane (TM) proteins

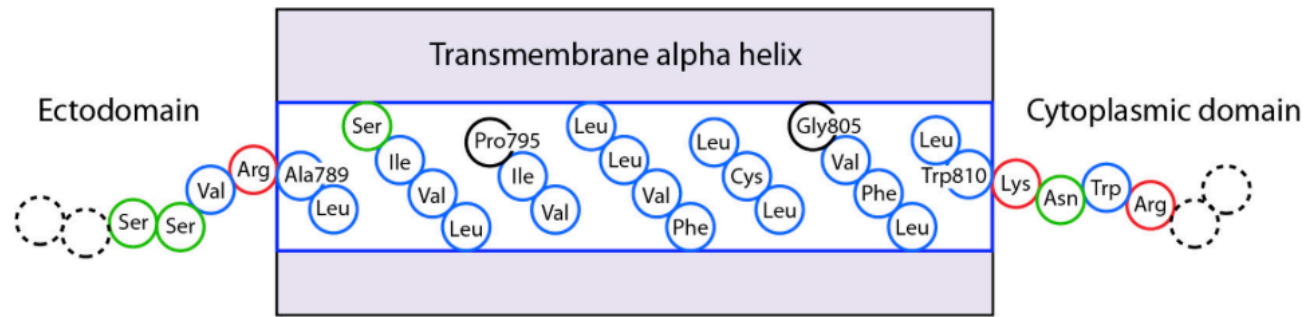
- Extremely difficult to solve membrane structures experimentally!
 - Only a few hundred structures in the PDB
- Can not use the same predictors for secondary structure as for globular proteins
- Special predictors for
 - helical membrane proteins
 - β -barrel proteins
- Pattern in TM α -helical proteins is:
 - 17-25 mainly hydrophobic TM helices
 - <60 residues polar connectors
- Predictions based on scanning for segments with high score for hydrophobicity
- Improved with neural networks

Tools:

- TMHMM
- Phobius



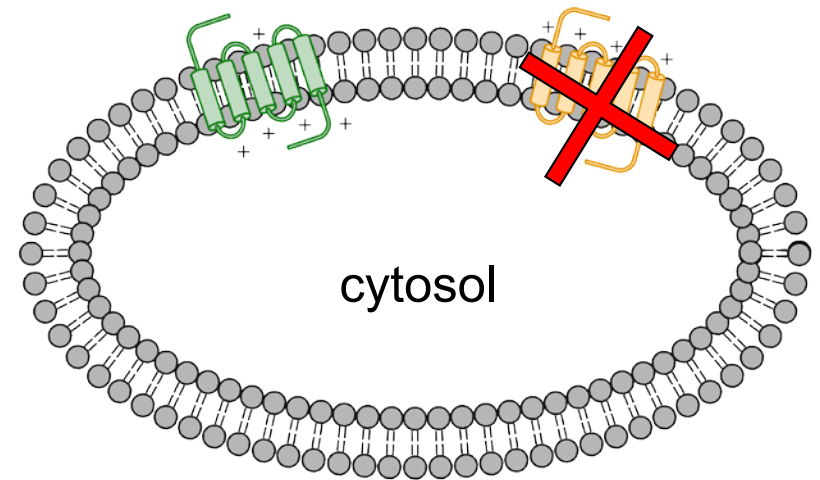
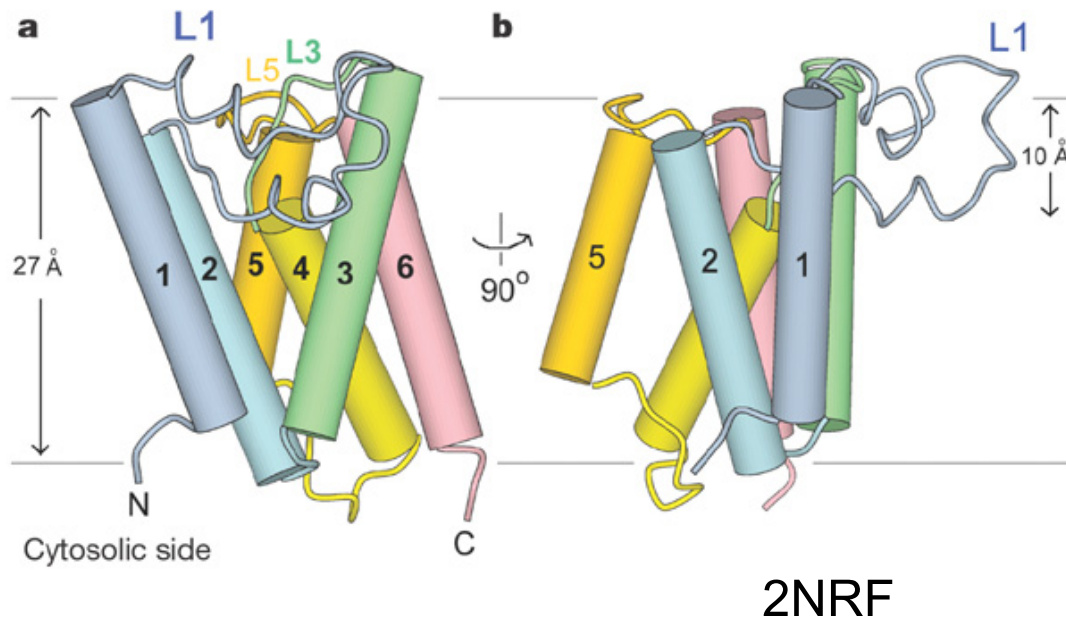
TM single-pass protein



Human	I	V	T	M	S	H	-	A	L	G	D	V	A	G	R	G	N	E	K	K	P	S	-	V	R	A	L	S	I	V	L	P	I	V	L	L	V	F	L	C	L	G	V	F	L	L	W	K	N	W	R	L	K	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C		
Macaque	T	V	T	M	S	H	-	A	L	G	D	V	A	G	R	G	N	E	K	K	P	K	S	-	V	G	A	L	S	I	V	L	P	I	V	L	L	V	F	L	C	L	G	A	F	L	L	W	K	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C	
Marmoset	I	V	T	M	S	H	-	A	L	G	D	V	A	G	R	G	T	E	E	K	P	R	S	-	V	G	A	L	S	I	L	P	I	V	L	L	V	F	L	C	V	G	A	F	L	L	W	R	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C		
Bushbaby	I	V	T	M	S	F	-	A	L	G	D	I	A	G	R	N	E	K	K	P	G	S	-	V	G	A	L	S	I	V	L	P	I	A	I	L	V	L	C	F	G	A	F	L	V	W	K	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C			
Mouse	S	V	T	V	S	H	-	V	Q	G	D	M	A	G	R	G	N	E	E	Q	P	H	G	-	M	R	F	L	S	I	F	F	P	I	A	L	V	A	L	L	V	L	G	A	V	L	L	W	R	N	W	R	L	K	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	L	H	I	C	
Rat	S	V	T	V	S	S	-	V	Q	G	D	M	A	G	R	G	D	E	V	Q	R	H	G	-	V	G	F	L	S	I	F	L	P	I	A	L	V	A	L	L	V	F	G	A	I	L	L	W	R	N	W	R	L	R	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	I	H	I	C	
Guinea pig	S	V	T	R	S	Q	-	-	V	A	D	A	A	G	R	G	D	-	-	K	P	R	G	-	V	G	A	L	T	I	A	L	P	I	G	L	L	T	L	L	C	L	G	A	F	L	V	W	K	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C	
Hamster	S	V	T	M	S	H	-	V	Q	G	D	-	-	-	-	R	R	N	E	E	R	P	Q	G	-	V	G	V	L	S	I	T	L	P	I	A	L	V	I	L	L	V	F	G	A	I	L	L	W	R	N	W	R	L	R	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	L	H	I	C
Squirrel	S	A	T	L	S	H	-	V	L	A	D	V	A	D	R	G	K	E	E	K	P	R	S	-	V	G	A	L	S	I	V	L	P	I	A	L	V	L	L	C	F	G	A	F	L	V	W	K	N	W	R	L	K	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C		
Horse	T	V	T	V	S	H	-	A	L	G	D	A	A	G	R	G	E	E	R	P	R	G	-	V	G	A	L	S	I	V	L	P	I	A	L	L	I	V	L	C	F	G	T	F	L	L	W	K	N	W	R	L	K	N	V	N	S	I	H	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C		
Cat	T	V	T	V	S	H	-	A	L	G	D	A	A	S	R	G	D	E	E	R	P	R	S	-	V	G	A	L	Y	V	I	L	P	I	V	L	L	I	L	L	G	F	G	T	F	L	L	W	K	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C	
Dog	M	V	T	M	S	H	-	A	L	G	D	A	A	S	R	G	D	-	E	K	P	R	S	-	V	G	A	L	Y	I	L	P	I	V	L	L	I	L	L	C	F	G	T	F	L	L	W	K	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C		
Panda	T	V	T	M	S	Q	H	-	A	L	G	D	A	A	S	R	G	D	E	E	K	P	R	S	-	V	G	A	L	Y	I	L	P	I	V	L	L	I	L	L	C	F	G	T	F	L	L	W	K	N	W	R	L	K	S	V	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C	
Pig	S	V	T	M	S	Q	H	-	A	L	G	D	V	A	G	R	G	V	T	E	K	P	Q	S	-	V	G	A	L	Y	I	V	L	P	I	A	L	L	I	L	L	F	F	G	T	F	L	L	W	K	N	W	R	L	K	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C
Dolphin	S	V	T	M	S	Q	-	V	L	G	D	V	A	G	R	A	D	A	E	R	P	Q	G	-	V	G	A	L	Y	I	V	L	P	I	A	L	L	I	L	L	C	F	G	T	F	L	L	W	K	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C	
Bovine	S	V	T	M	S	Q	-	G	Q	G	D	I	A	S	Q	A	D	T	E	R	P	G	S	-	V	G	A	L	Y	I	V	L	P	I	A	L	L	I	L	L	A	F	G	T	F	L	L	W	K	N	W	R	L	K	S	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C	
Tasmanian devil	M	V	T	M	S	Q	K	-	A	V	Q	N	P	M	G	M	Q	D	T	P	E	H	Q	G	G	S	K	A	L	I	V	L	P	I	V	L	I	S	L	I	C	F	G	A	Y	V	V	W	K	K	W	R	L	R	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C	
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Rattlesnake	M	V	T	L	S	Q	-	V	Q	N	G	I	A	V	E	T	D	G	T	E	R	R	G	-	P	S	A	L	A	I	V	L	P	L	A	L	I	S	L	V	S	F	G	T	F	L	I	W	K	N	W	R	L	K	N	I	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	I	H	I	C	
Anole lizard	M	V	T	L	S	Q	-	A	Q	N	G	I	A	A	E	I	D	G	A	K	Q	R	G	-	P	P	A	L	S	I	V	L	P	L	V	L	I	C	L	V	S	F	G	A	Y	L	V	W	K	N	W	R	L	K	N	T	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	I	H	I	C	
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Coelacanth	P	K	V	L	T	N	E	-	A	Q	T	G	V	A	A	E	M	T	T	E	H	H	G	-	-	P	T	A	L	Y	I	V	L	P	I	V	I	L	S	L	L	C	F	G	A	F	L	L	W	R	N	W	K	L	K	N	T	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C
Shark	S	T	V	P	T	L	H	-	E	L	L	T	V	S	A	K	A	A	T	E	G	H	R	G	-	T	N	A	L	W	I	V	L	P	L	A	I	L	S	L	L	A	T	A	T	Y	F	I	W	K	N	W	K	L	K	N	T	N	S	I	N	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	T
Platyfish 1	P	A	G	T	-	-	-	-	P	Q	G	F	Q	K	A	V	M	P	E	E	A	P	A	S	H	S	V	A	L	Y	V	F	L	P	L	G	I	I	A	A	L	V	C	G	A	V	L	F	W	R	N	W	H	R	K	N	T	N	T	I	H	F	A	N	P	V	Y	Q	K	T	T	-	E	D	E	V	H	I	C
Stickleback 1	G	T	V	F	S	A	V	-	S	G	N	R	L	A	A	S	P	E	E	A	H	L	S	H	-	P	V	A	L	Y	V	V	L	P	I	L	V	M	S	L	L	V	F	G	A	M	W	V	W	R	H	W	R	L	K	N	T	N	T	I	H	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	M	H	I	C
Tilapia 1	M	L	T	F	S	K	V	-	G	D	R	Q	A	A	V	Q	P	E	E	A	P	S	S	H	P	V	A	L	Y	V	V	L	P	L	M	I	M	A	L	L	C	F	G	A	F	L	L	W	R	N	W	K	L	K	N	T	N	T	I	H	F	D	N	P	V	Y	Q	K	T	T	-	E	D	E	L	H	I	C	
Platyfish 2	P	V	T	T	P	G	V	-	P	P	K	G	V	A	V	I	P	E	T	T	S	T	-	-	S	K	T	L	Y	F	V	L	P	L	A	I	L	C	L	V	A	V	G	G	V	L	L	W	R	Y	R	L	Q	N	T	N	T	M	H	F	N	P	V	Y	Q	K	T	T	-	E	D	O	V	H	I	W			
Stickleback 2	P	V	T	S	P	D	F	-	K	H	R	F	V	A	A	I	P	S	T	A	S	P	A	-	-	P	I	S	L	Y	I	A	L	P	L	A	V	C	L	V	A	G	G	V	L	L	W	R	N	Y	R	L	K	N	T	N	T	I	H	F	D	N	P	V	Y	Q	K	T	T	-	E	D	O	V	H	I	Y		
Tilapia 2	P	V	T	S	P	N	-	-	I	R	P	E	F	T	E	V	L	T	K	T	V	S	T	-	-	P	I	A	L	Y	V	I	L	P	L	A	V	A	L	V	A	V	G	G	V	L	L	W	R	N	Y	R	L	K	N																								

Transmembrane (TM) proteins – Secondary structure prediction

- Prediction of membrane orientation (in-out)
- *Positive-inside rule*: Residues at cytosolic side are more positively charged than at the luminal/periplasmic side

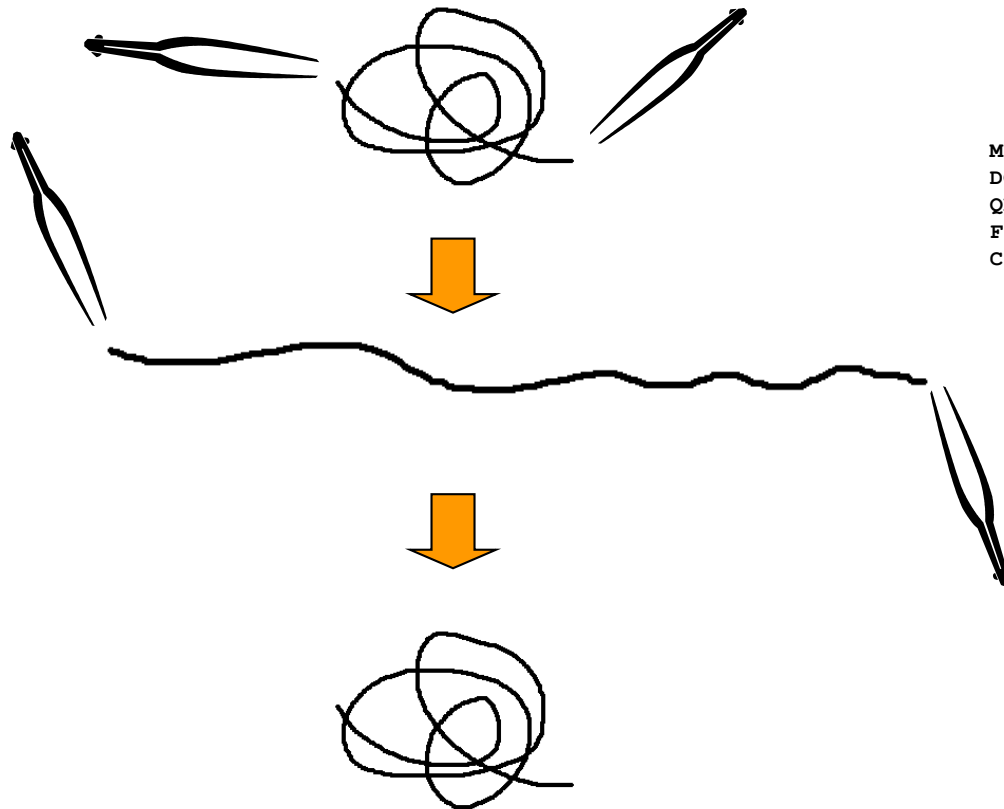
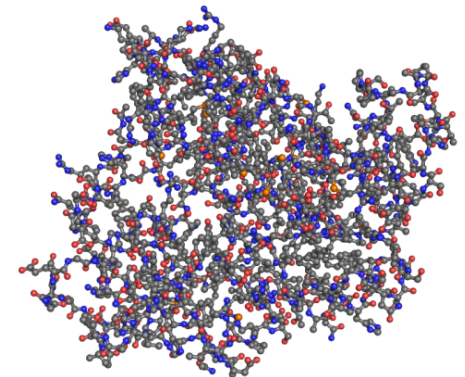


3D structure modeling

Modeling of 3D structure

Jon K. Lærdahl,
Structural Bioinformatics

- ~135,000,000 sequence records in the traditional GenBank divisions (Apr 2011)
 - Several orders of magnitude more sequences in other public databases
 - Next Generation Sequencing generates ~20 Gb in *a single run*
- ~135,000 3D structures in the PDB (*i.e.* all published structures)
 - Solving a single structure experimentally takes 1-3 yrs
 - Some protein structures are “close to impossible” to solve, *e.g.* many membrane proteins
- In the cell, the sequence determines the 3D structure of the protein



Folding is spontaneous
in the cell (but often
with helper molecules,
chaperones)

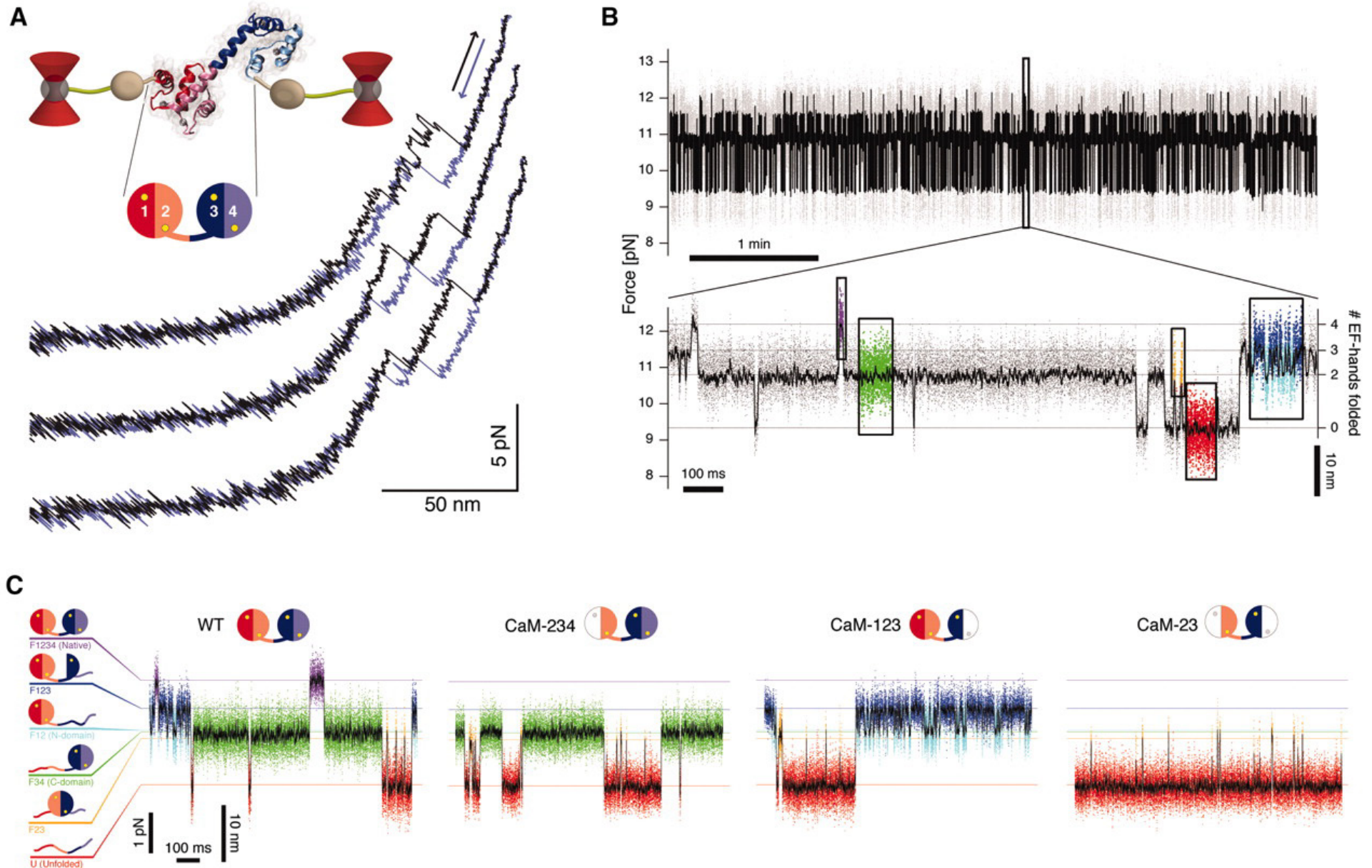
```
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DQVWTLTQTEELHCTVYRGDKSQASRPTPDELEAVRKYFQLDVTLAQLYHHWGSVDSHF  
QEVAQKFQGVRLLRQDPICLFSFICSSNNNIARITGMVERLCQAFGPRLIQLDDVTYHG  
FPSLQALAGPEVEAHLRKLGLGYRARYVSASARAILEEQGGLAWLQQLRESSYEEAHKAL  
CILPGVGTKVADCIICLMALDKPQAVPVDVHMWHIAQRDYSWHPTTSQAKGPSPTNKELG
```

The sequence
determines the 3D
structure!

Nobel Prize in chemistry
1972 to Christian B.
Anfinsen

Optical tweezers

Jon K. Lærdahl,
Structural Bioinformatics



Stigler *et al.*, Science **334**, 512 (2011).

Protein folding

MPARALLPRRMGHRTLASTPALWASIPCPRSELRLDLVLP
SGQSFRWREQSPAHWSGVLA
DQVWTLTQTTEEQLHCTVYRGDKSQASRPTPDELEAVRKYFQ
LDVTLAQLYHHWGSVDSHF
QEVAQKFQGVRLLRQDPIECLFSFICSSNNNIARITGMVERLCQAF
GPRLIQLDDEVITYHG
FPSLQALAGPEVEAHLRKLGLGYRARYVSASARAILEEQGGLAWLQQL
RESSYEEAHKAL
CILPGVGTKVADCICLMALDKPQAVPVDVHMWHIAQRDYSWHPTTSQAKG
PSPQTNKELG
NFFRSLWGPYAGWAQATPPSYRCCSVPTCANPAMLRSHQQSAERV
PKGRKARWGTLDEI

Folding is
spontaneous in the
cell

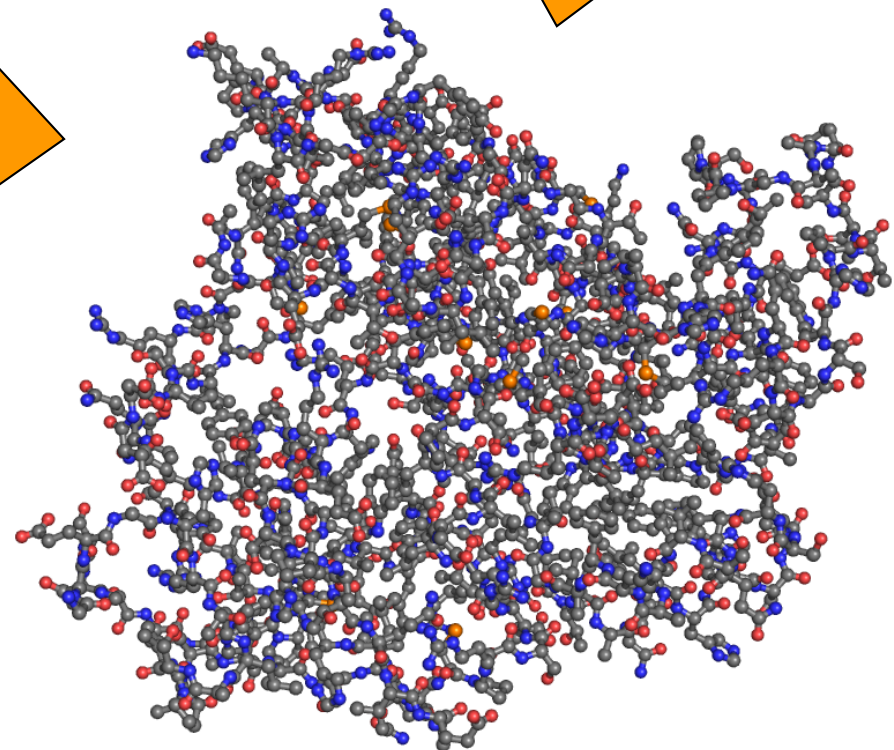
The sequence determines
the 3D structure!

In the cell

In the computer

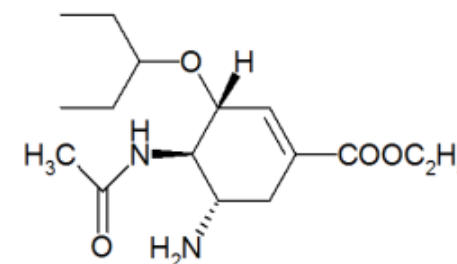
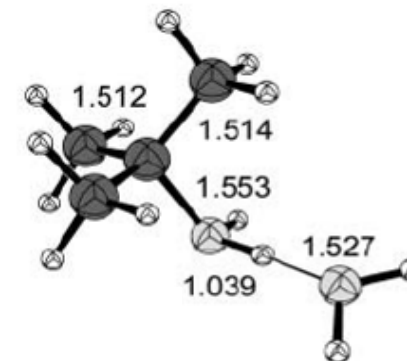
***Ab initio/de novo* structure prediction**

- Based on physical/chemical laws
and not already published
experimental structures



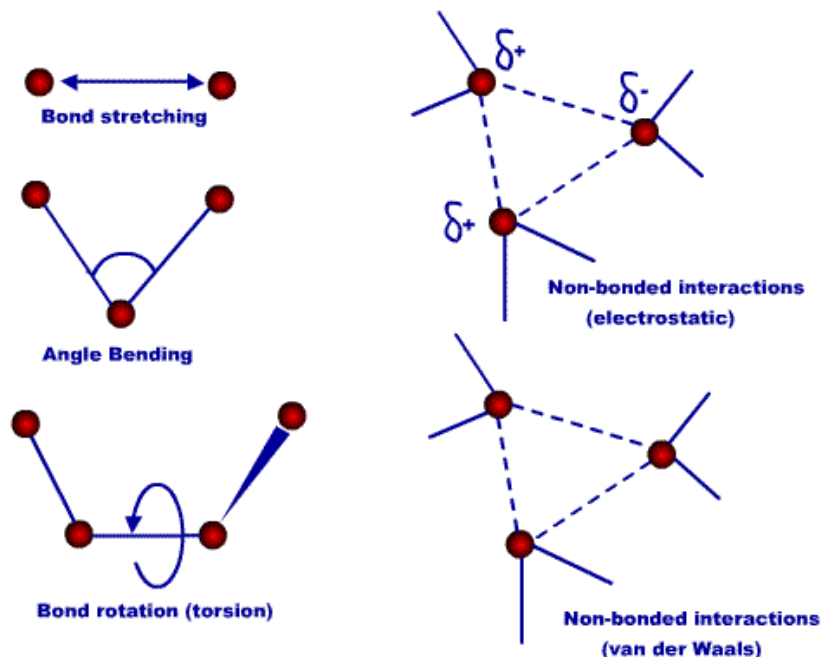
Ab initio structural prediction

- Determine the tertiary structure for a protein based on amino acid sequence and chemical and physical laws only
- Does *not* use prior knowledge of structure from the PDB
- *Ab initio* quantum chemistry is pure “*ab initio*”
 - Based on solving the Schrödinger equation
 - Is routinely used for chemical systems of up to 20-50 atoms
 - Can be used to compute/model the correct 3D structure for drug candidates, small metabolites or tiny peptides
 - Will *not soon* be applicable for large proteins with 1000s of atoms
- *Ab initio* protein 3D structure prediction
 - Also called *de novo* structure prediction/protein modeling
 - Is *not* based on solving the Schrödinger equation
 - Instead uses more approximate methods for energy minimization/folding (Confusing: This is exactly what is *not ab initio* quantum chemistry)
 - Extremely computationally intensive
 - Very hard! This field is far from mature...
 - Only possible (useful/reliable) for small (poly)peptides (less than 10-100 residues?)



Ab initio structural prediction

- Molecular mechanics/force field calculations – Newtonian mechanics to model proteins
 - Each atom simulated as a single particle
 - Each particle has a size (van der Waals radius), charge and polarizability
 - Bonded interactions are treated as “springs” with a given equilibrium bond distance – same for bond angles and dihedral angles
 - Additional terms, e.g. non-bonded collisions, solvent etc.



$$\begin{aligned}
 U(\vec{R}) = & \sum_{\text{bonds}} K_b (b - b_0)^2 + \sum_{\text{angles}} K_\theta (\theta - \theta_0)^2 \\
 & + \sum_{\text{Urey-Bradley}} K_{UB} (S - S_0)^2 \\
 & + \sum_{\text{dihedrals}} K_\phi (1 + \cos(n\phi - \delta)) + \sum_{\text{impropers}} K_\omega (\omega - \omega_0)^2 \\
 & + \sum_{\text{non-bonded pairs}} \left\{ \epsilon_{ij}^{\min} \left[\left(\frac{R_{ij}^{\min}}{r_{ij}} \right)^{12} - 2 \left(\frac{R_{ij}^{\min}}{r_{ij}} \right)^6 \right] + \frac{q_i q_j}{4\pi\epsilon_0 \epsilon r_{ij}} \right\} \\
 & + \sum_{\text{residues}} U_{\text{CMAP}}(\phi, \psi)
 \end{aligned}$$

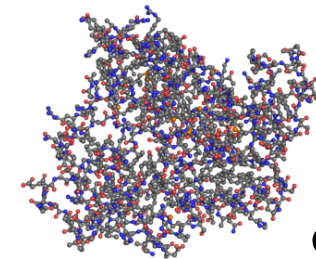
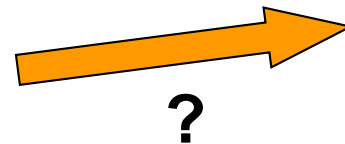
Brooks *et al.*, J. Comput. Chem. **30**, 1545 (2009).

Ab initio structural prediction

Jon K. Lærdahl,
Structural Bioinformatics

- Does *not* use prior knowledge of structure from the PDB
 - That is why they are known as *ab initio*
- Still, some programs known as *ab initio* protein modeling programs also use *some* information from the PDB, for example structures for small fragments
- At least in some respects based on the “paradigm” of Anfinsen that all information that is needed to determine the tertiary structure is in the primary sequence
 - Is it really correct?
 - Certainly not always!
 - Folding chaperons
 - Ribosomal environment, timing of protein synthesis, solvent, salinity, pH, temperature, metabolites and other macromolecules, etc. may (and do) in many cases contribute to the folding process
- All problems with *ab initio* modeling will never be completely solved?
- They have certainly not been solved yet!

```
MPARALLPRRMGHRTLASTPALWASIPCPRSELRLDLVLPSCQSFRWREQSPAHWSGVLA  
DQVWTLTQTEELHCTVYRGDKSQASRPTPDELEAVRKYFQLDVTLAQLYHHWGSVDSHF  
QEVAQKFQGVRLRQDPIECLEFSFICSSNNNIARITGMVERLCQAFGPRLIQLDVITYHG  
FPSLQALAGPEVEAHLRLGLGYRARYVSASARAILEEQGLAWLQQLRESSYEEAHKAL  
CILPGVGTKVADCI CLMALDKPQAVPVDVHMWHIAQRDYSWHPTTSQAKGPSPQTNKELG
```



or



Robetta: full-chain protein structure prediction server - Windows Internet Explorer

http://robetta.bakerlab.org/

Robetta: full-chain protein structure prediction server

www.bakerlab.org

ROBETTA

Full-chain Protein Structure Prediction Server

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SERVICES

Domain Parsing & 3-D Modeling
(homology modeling, *ab initio* structure prediction, and structure prediction using NMR constraints)
[\[Queue \]](#) [\[Submit \]](#)

Interface Alanine Scanning
[\[Queue \]](#) [\[Submit \]](#)

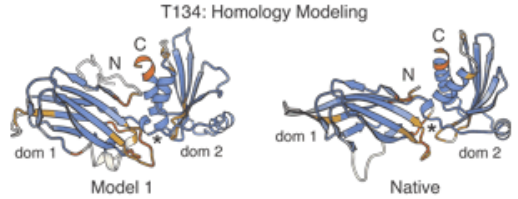
Fragment Libraries
[\[Queue \]](#) [\[Submit \]](#)

DNA Interface Amino Acid Affinity/Specificity Scan
[\[Queue \]](#) [\[Submit \]](#)

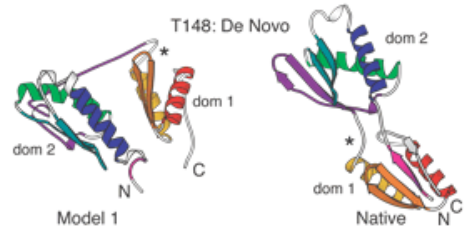
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T134: Homology Modeling



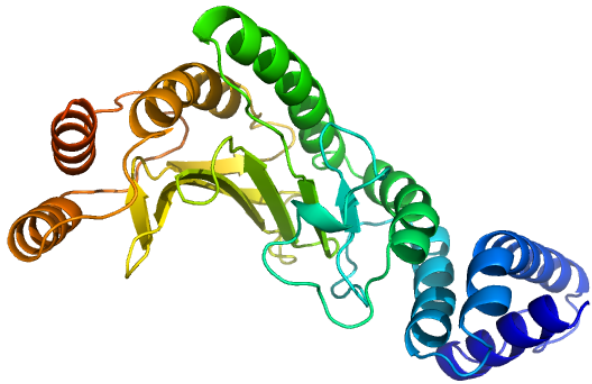
T148: De Novo



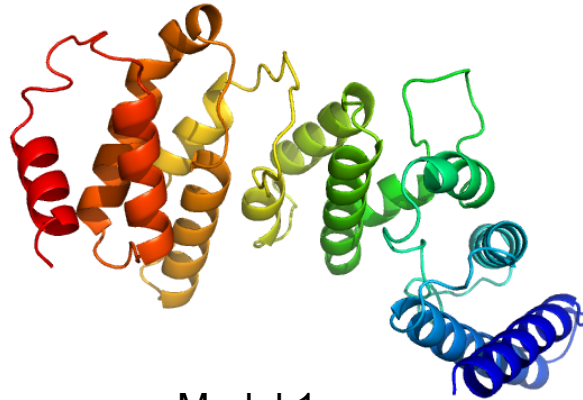
examples of predictions by Robetta in CASP-5

David Baker

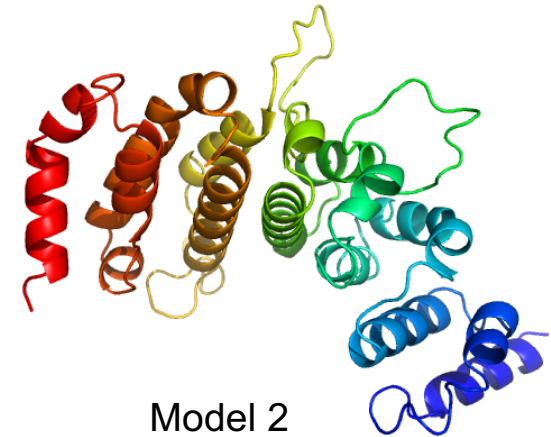
I-TASSER from Yang Zhang-lab is another possibility. Ranked as no. 1 in "structure prediction competition" in 2006, 2008, 2012, and 2014 (Actually not pure *ab initio*).



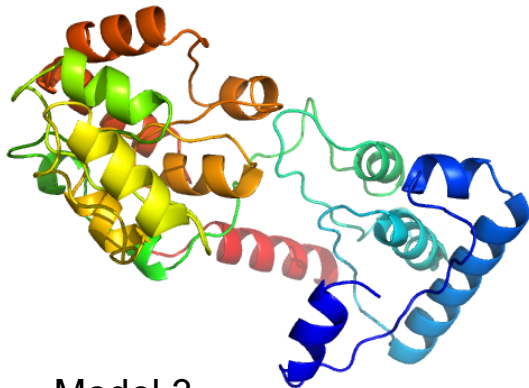
Experimental 3D structure of my colleague



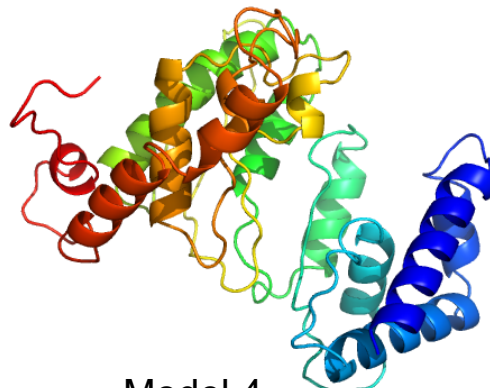
Model 1



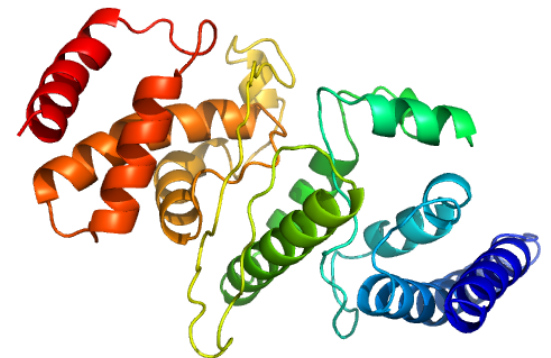
Model 2



Model 3



Model 4



Model 5

3D structure modeling

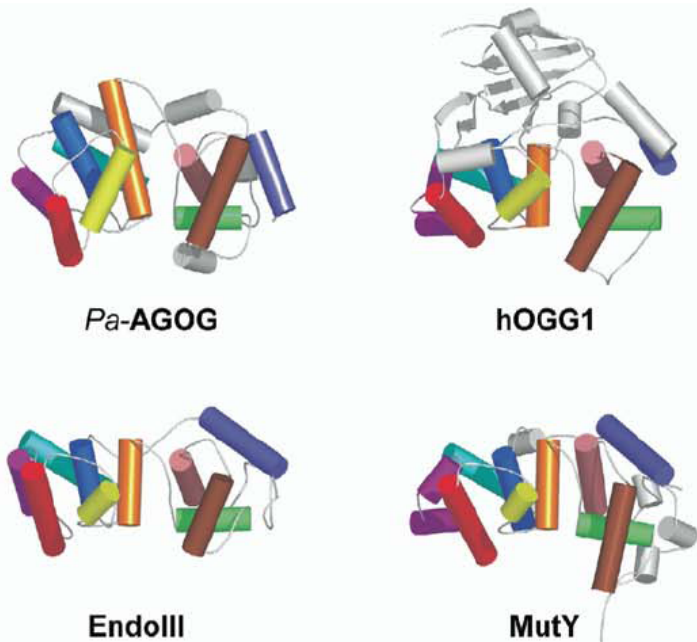
- *Ab initio/de novo* – very hard...
- Threading/fold recognition
- Homology modeling

Protein structure evolution

Jon K. Lærdahl,
Structural Bioinformatics

```

OGG1_YEAST/1-376      174 SRATEAKLRELGFGRARYI IETARKLVNDKAEANITSDTTYLQSICKDAQYEDVREHLMSYNGVGPKVADCVCLMGLHMDGIVPVDVHVSRIAKRDYQISAN 276
OGG1_MOUSE/1-345     189 GPEAETHLRKLGGLGYRARIYRASAKAILEEQGGP-----AWLQQLRV-APYEEAHKALCTLPGVGAKVADCICLMALDKPOAVPVDVHWWQIAHRDYGWHPK 284
OGG1_RAT/1-345       189 GPEVETHLRKLGGLGYRARIYCASAKAILEEQGGP-----AWLQQLRV-ASYEEAHKALCTLPGVGAKVADCICLMALDKPOAVPVDIHWWQIAHRDYGWHPK 284
OGG1_HUMAN/1-345     189 GPEVEAHLRKLGLGYRARIYVSASAKAILEEQGGL-----AWLQQLRE-SSYEEAHKALC ILPGVGTQVADCICLMALDKPOAVPVDVHMHIAQRDYSWHP 284
OGG1_FLY/1-343       191 CEDLNAQLRAAKFGYRAFI AQTLQEI QKKGGQ-----NWFISLKS-MPF EKAREEL TLLPGIGYKVADCICLM SMGHLESVPVDIH IYR I AQNYYLPHLT 285
    
```



- Reason for similarities in sequence/structure is **common ancestry**, the sequences/structures are **homologs**
- Structures evolves slowly
- Sequence evolves faster
 - Many mutations does not change the structure
- Only some few 1000 superfamilies in the PDB
- Only a factor 2-10(???) as many superfamilies in Nature? Some few 1000 folds?

SCOP

Class	Number of folds	Number of superfamilies	Number of families
All alpha proteins	259	459	772
All beta proteins	165	331	679
Alpha and beta proteins (a/b)	141	232	736
Alpha and beta proteins (a+b)	334	488	897
Multi-domain proteins	53	53	74
Membrane and cell surface proteins	50	92	104
Small proteins	85	122	202
Total	1086	1777	3464

CATH

Class	Architecture	Topology	Homologous Superfamily
1	5	310	682
2	20	196	438
3	14	512	956
4	1	92	102
Total	40	1110	2178