



Proteins: Structure & Function

Ulf Leser

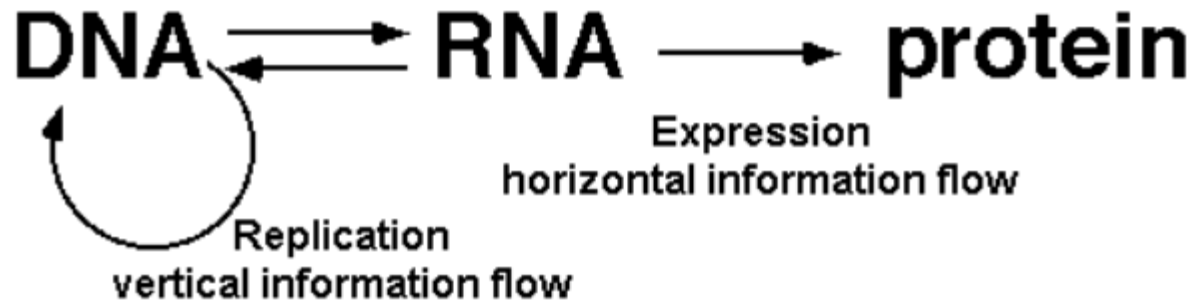
Praktikum / Abschlussarbeit bei MicroDiscovery

- Thema: Erstellung einer Android App zur Geräteprüfung
- Wir suchen einen Praktikanten mit Interesse an einem praxisnahen Projekt im Bereich Softwareentwicklung. Im Rahmen der Produktion eines mobilen Reader Gerätes soll eine App entwickelt werden. Die App soll firmenintern in der Qualitätssicherung zum Einsatz kommen und unterstützt die Fertigung einer Baugruppe. Die vorgesehenen Arbeiten umfassen dabei viele für die Softwareentwicklung relevante Bereiche wie:
Entwurf und Design der App
 - Algorithmenentwicklung für die spezifischen Fragestellungen
 - Überlegungen zur Useability
 - Implementation der Algorithmen und Klassenstrukturen
 - Testen (Unit-Tests, GUI-Tests, Performance-Tests)
 - Validierung der App in einer kleinen Studie
- Das Praktikum ist auf mindestens 2 Monate ausgelegt. Sie sollten Interesse an der Arbeit in einem Softwareunternehmen haben und Erfahrung in folgenden Bereichen mitbringen: Java und Android Programmierung (mit der Eclipse Entwicklungsumgebung)

This Lecture

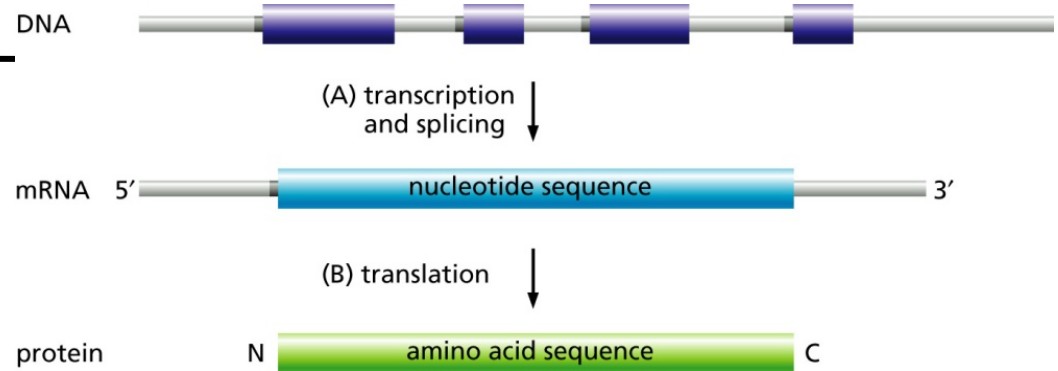
- Proteins
 - Structure
 - Function
 - Databases
- Predicting Protein Secondary Structure
- Many figures from Zvelebil, M. and Baum, J. O. (2008). "Understanding Bioinformatics", Garland Science, Taylor & Francis Group.
- Examples often from O. Kohlbacher, Vorlesung Strukturvorhersage, WS 2004/2005, Universität Tübingen

Central Dogma of Molecular Biology



	U		C		A		G	
U	UUU Phenyl- alanine UUA Leucine UUG		UCU Serine UCC UCA UCG		UAU Tyrosine UAC UAA Stop codon UAG Stop codon		UGU Cysteine UGC UGA Stop codon UGG Tryptophan	U C A G
C	CUU Leucine CUC CUA CUG		CCU Proline CCC CCA CCG		CAU Histidine CAC CAA Glutamine CAG		CGU Arginine CGC CGA CGG	U C A G
A	AUU Isoleucine AUC AUA Methionine; initiation codon AUG		ACU Threonine ACC ACA ACG		AAU Asparagine AAC AAA Lysine AAG		AGU Serine AGC AGA Arginine AGG	U C A G
G	GUU Valine GUC GUA GUG		GCU Alanine GCC GCA GCG		GAU Aspartic acid GAC GAA Glutamic acid GAG		GGU Glycine GGC GGA GGG	U C A G

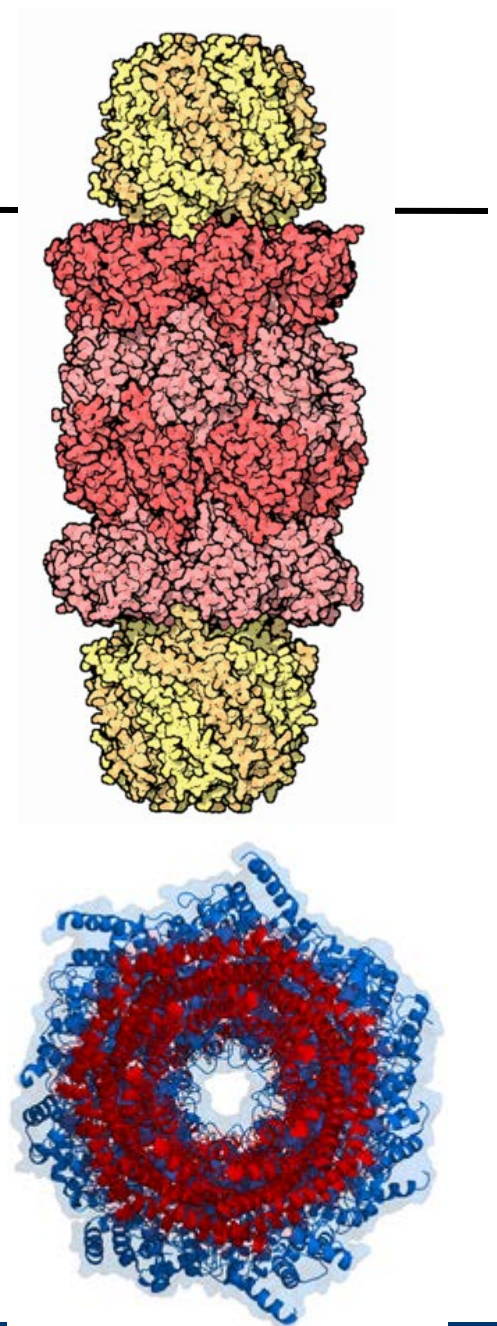
Details



- Alternative Splicing
 - “One gene – one protein” is wrong
 - Exons may be spliced out from the mRNA
 - Human: at least **6 times more unique proteins** than genes
- Post-translational modifications
 - (De-)Phosphorylation, glycolysation, cleavage of signal peptides, ...
 - Human: At least **5 times more protein forms** than proteins
- Complexes: Proteins **physically group together** to perform specific function

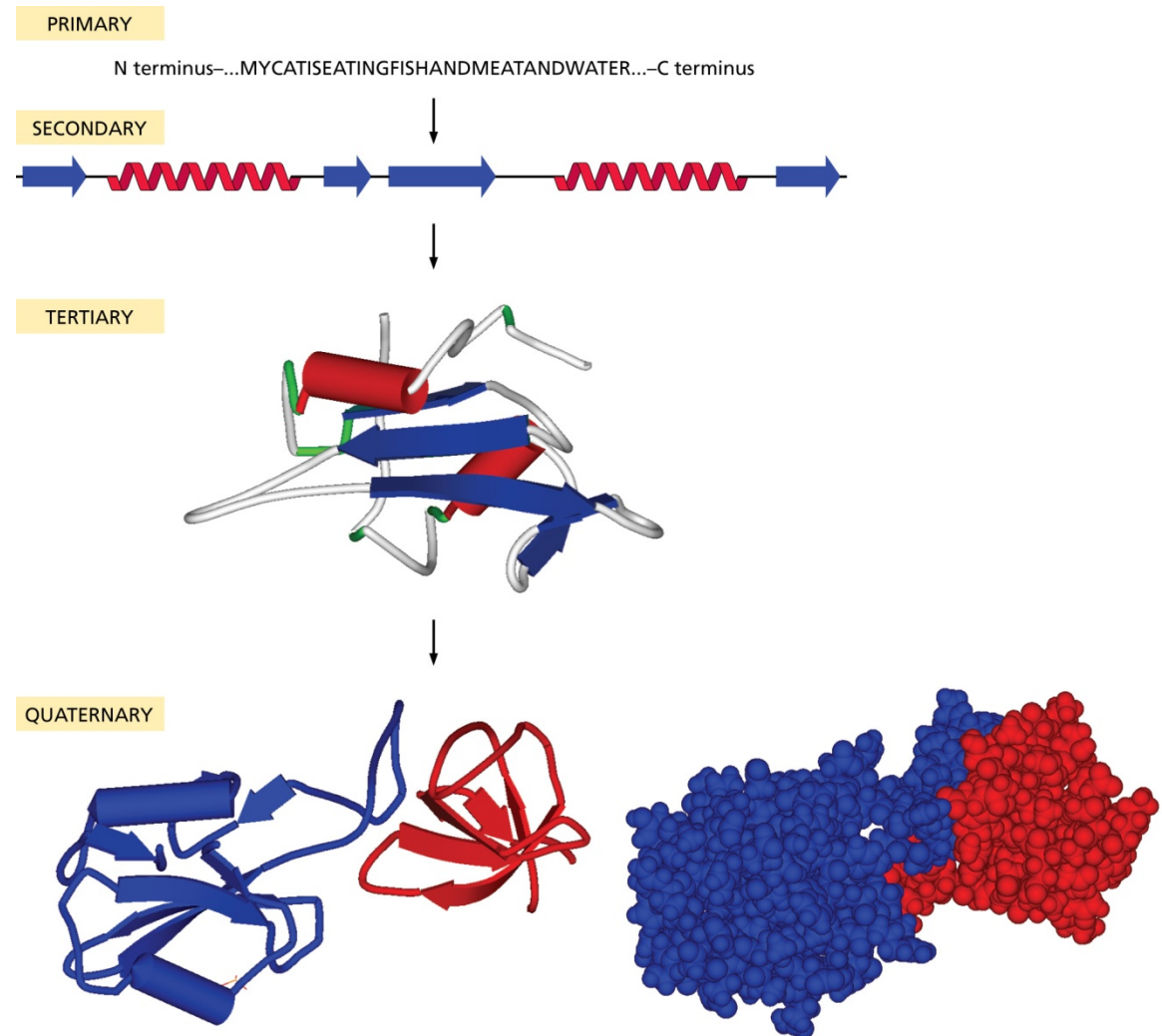
Example: Proteasome

- Function: Breaks (mis-folded, broken, superfluous, ...) proteins into small **peptides for reuse**
- Very large complexes present in all eukaryotes (and more species)
 - >2000 kDa, made of **dozens of single proteins**
 - Formation of the complex is a complex process only partly understood yet



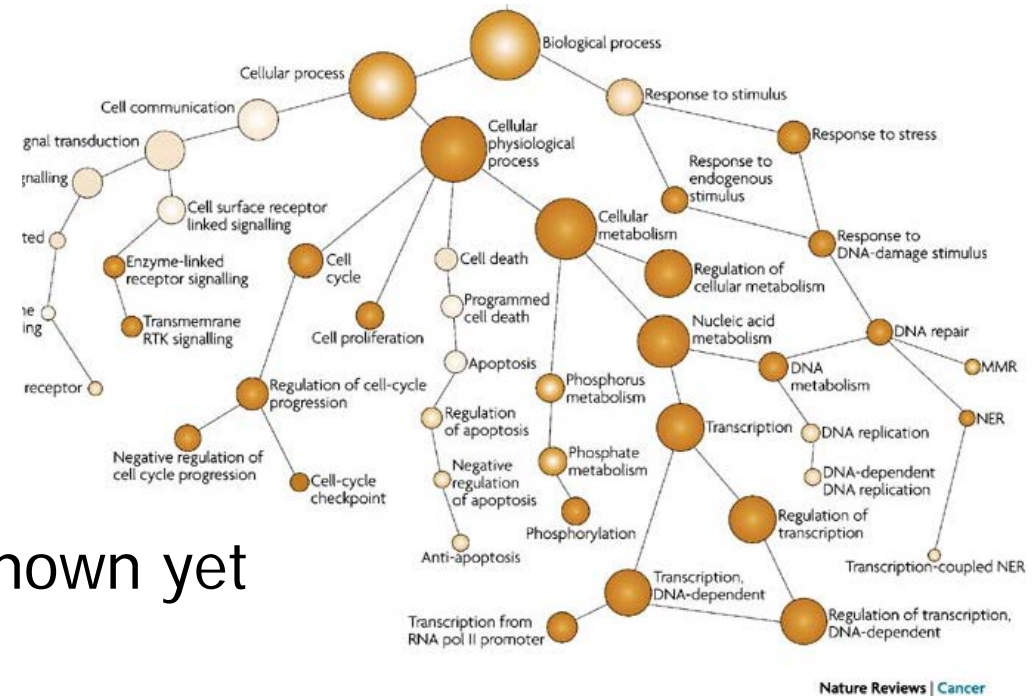
Protein Structure

- Primary
 - 1D-Seq. of AA
- Secondary
 - 1D-Seq. of “subfolds”
- Tertiary
 - 3D-Structure
- Quaternary
 - Assembled complexes

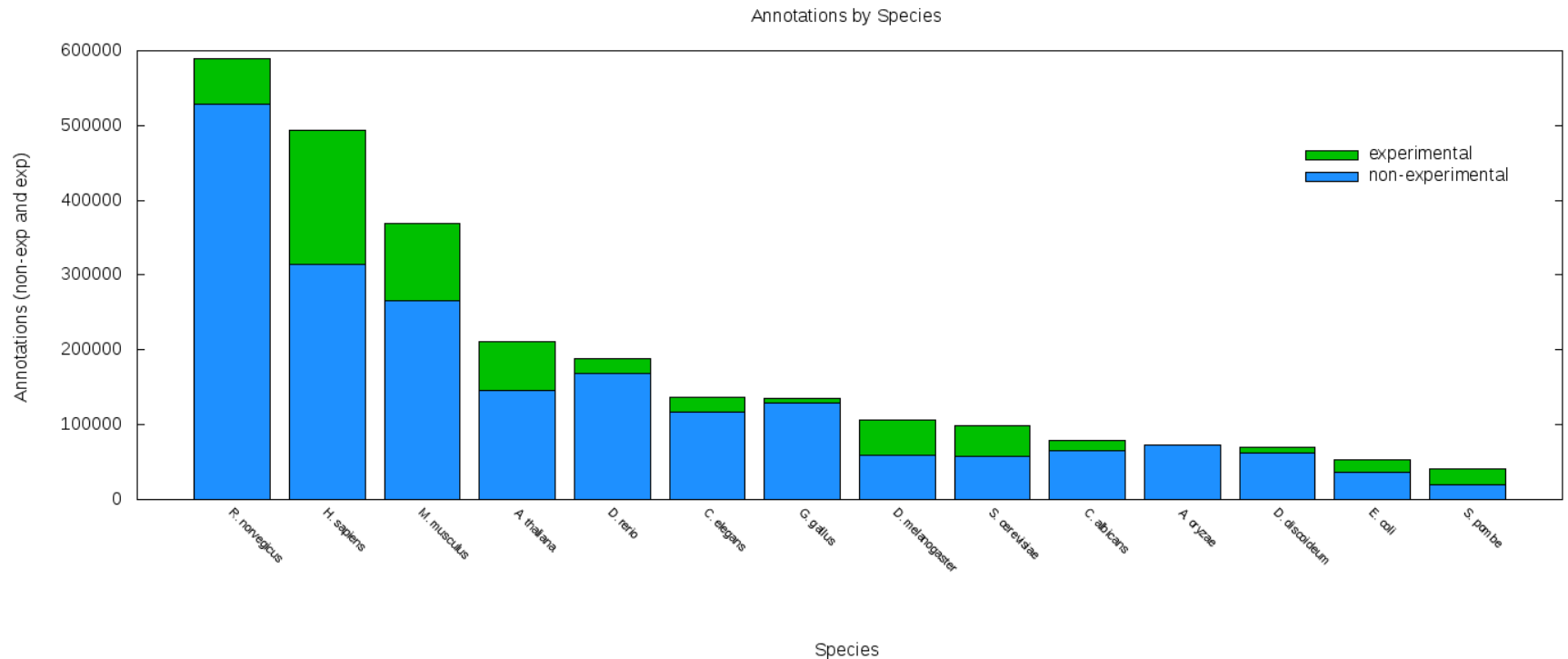


Protein Function

- Proteins perform essentially everything that makes an organism alive
 - Metabolism
 - Signal processing
 - Gene regulation
 - Cell cycle
 - ...
- For ~1/4 of all human gene, no function is known yet
- Describing function
 - **Gene Ontology**: 3 branches, >30.000 concepts
 - Used world-wide to describe gene/protein function

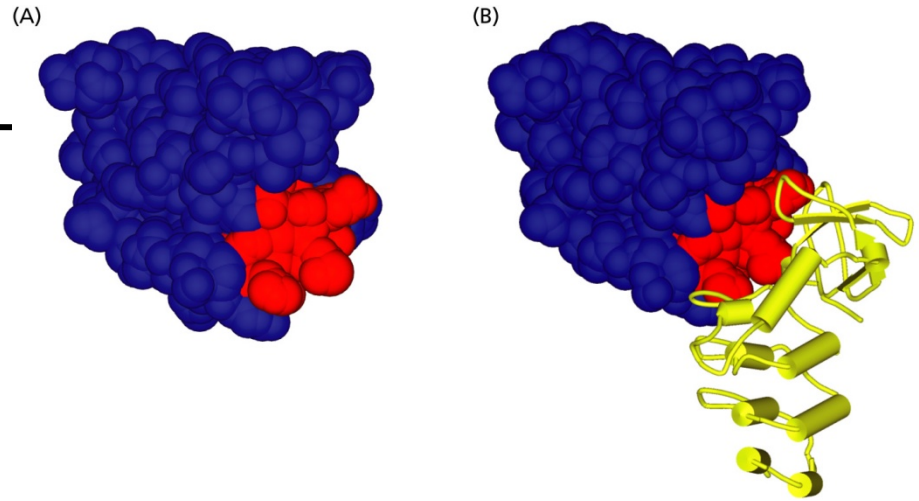


„Known“ Protein Functions



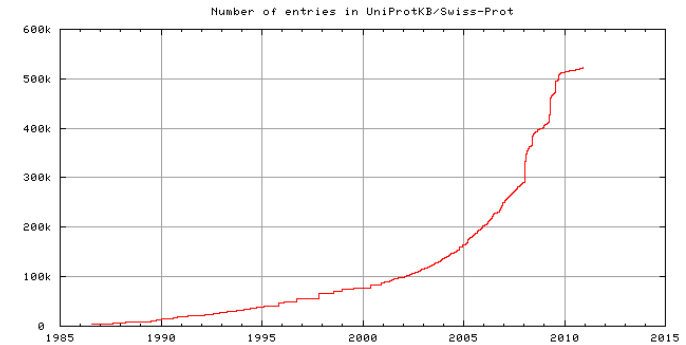
<http://geneontology.org/page/current-go-statistics>, June 2016

Function and Motifs

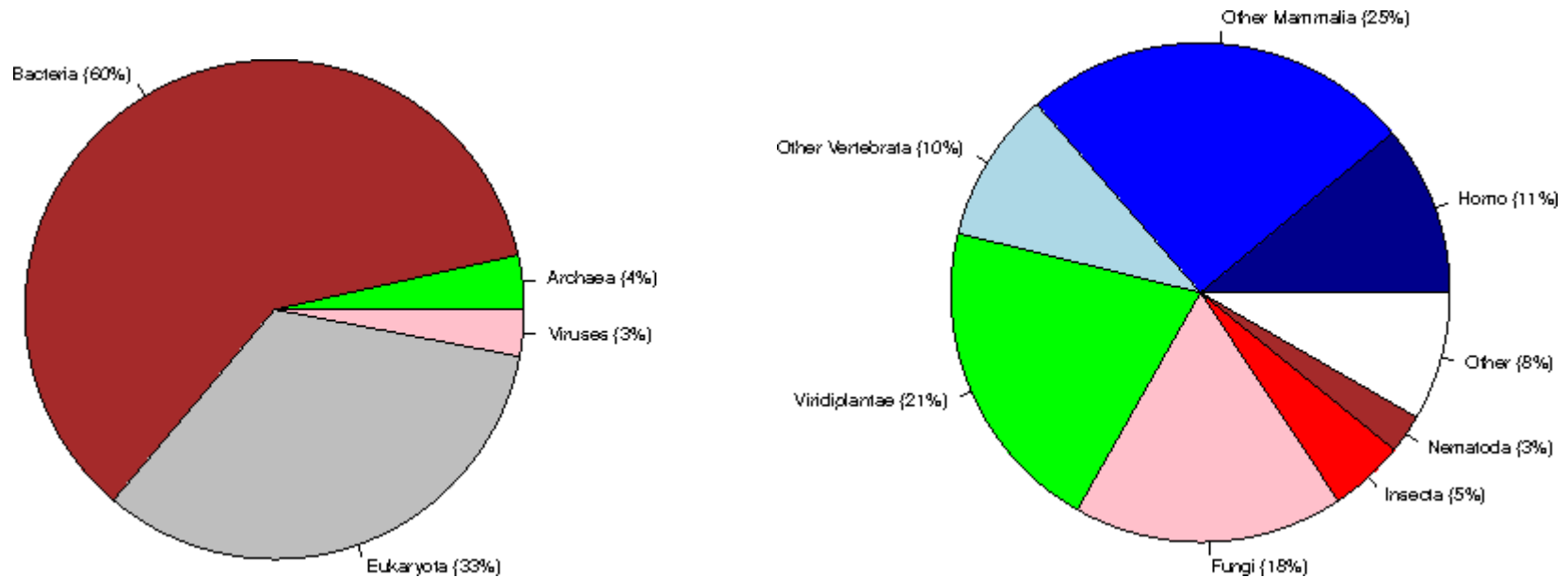


- Protein often have multiple functions
 - Avg. n# of GO terms assigned to a human protein: 6-10
- Functions are associated to **motifs or domains**
- There probably exist only 4000-5000 motifs
 - Proteins as assemblies of functional motifs
- Performing a function often requires **binding to another protein** or molecule
 - The binding requires a certain constellation of the protein structure
 - Major target of **pharmacological research**

- “Standard” database for **protein sequences and annotation**
 - Original name: SwissProt
 - Started at the Swiss Institute of Bioinformatics, now mostly EBI
 - Other: PIR, HPRD
- Continuous growth and **curation**
 - >30 „Scientific Database Curators”
 - Quarterly releases
 - **Very rich set of annotations**
- Actually two databases
 - **SwissProt**: Curated, high quality, versioned
 - TrEMBL: Automatic generation from (putative) coding genomic sequences, low quality, redundant, much larger



UniProt: Species [<http://www.expasy.org/sprot/relnotes/relstat.html>, June 2016]

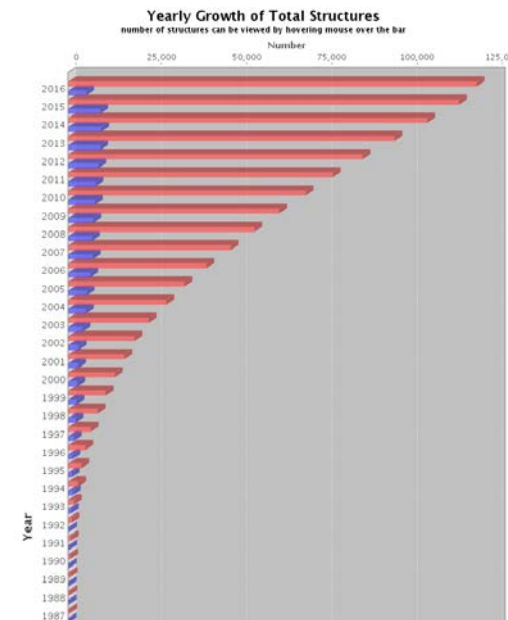


20258	Homo sapiens (Human)
16327	Mus musculus (Mouse)
9842	Arabidopsis thaliana (Mouse-ear cress)
7560	Rattus norvegicus (Rat)
6582	Saccharomyces cerevisiae (Baker's yeast)
5803	Bos taurus (Bovine)

...

PDB – Protein Structure Database

- Oldest protein database, evolved from a book
- Contains experimentally obtained **protein 3D-structures**
 - Plus some DNA, protein-ligand, complexes, ...
 - X-Ray (~75%), NMR (nuclear magnetic resonance, ~23%)
- Costly and **rather slow techniques**
 - Growth much smaller than that of sequence-related DBs
- Many problems with **legacy data** and data formats



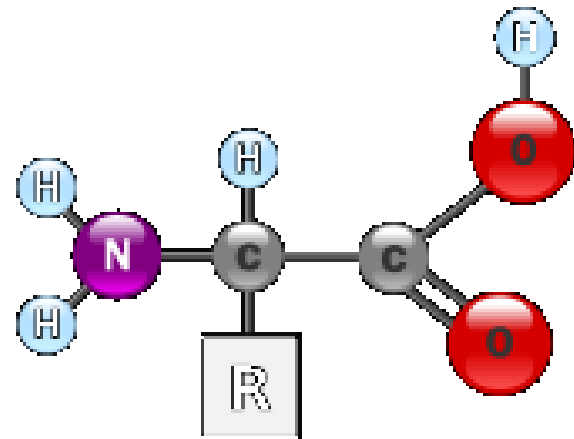
<http://www.rcsb.org/pdb/statistics/contentGrowthChart.do?content=total>, June 2016

This Lecture

- Introduction
- Predicting Protein Secondary Structure
 - Secondary structure elements
 - Chou-Fasman
 - GOR IV
 - Other methods

Amino Acids (AA)

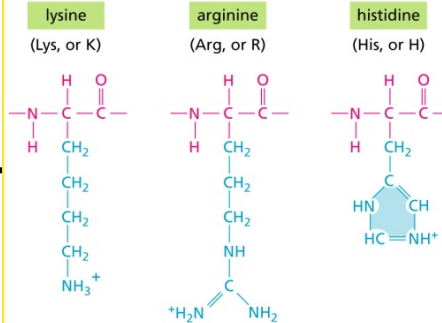
- AA consist of a common core and a **specific residue**
 - Amino group – NH_2
 - Central C_α - Carbon – CH
 - Carboxyl group – COOH



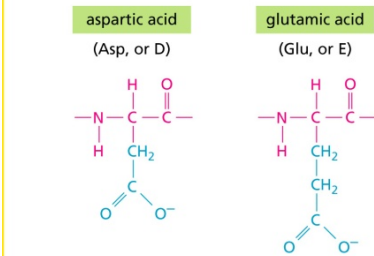
- Residues (side chains) vary greatly between AA
- Residues determine the **specific properties** of a AA

Side Chains

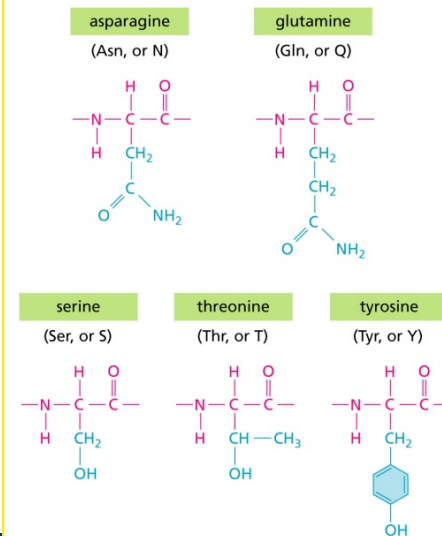
BASIC SIDE CHAINS



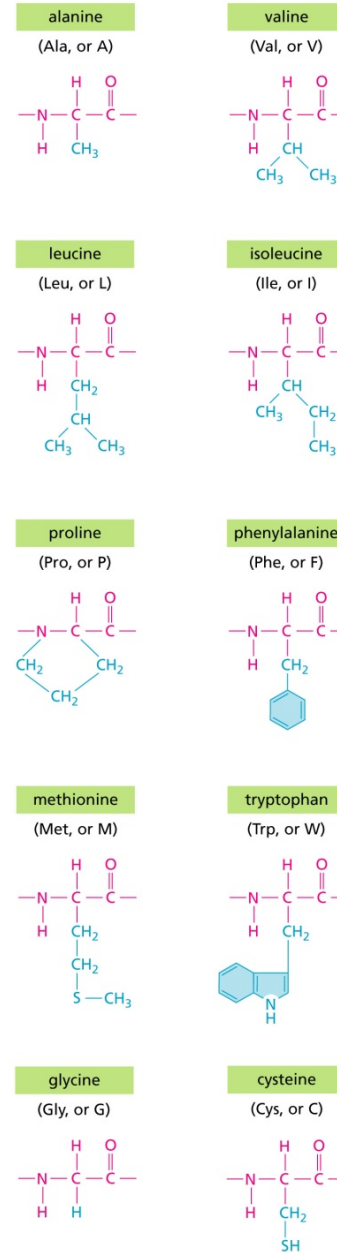
ACIDIC SIDE CHAINS



UNCHARGED POLAR SIDE CHAINS



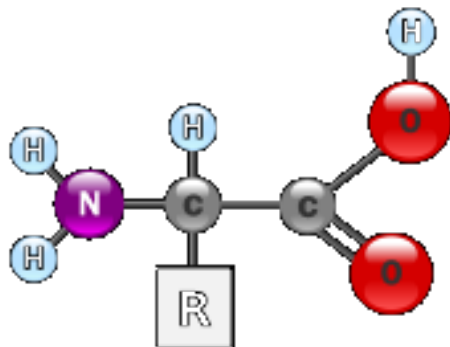
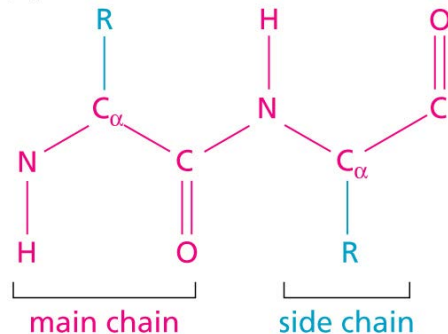
NONPOLAR SIDE CHAINS



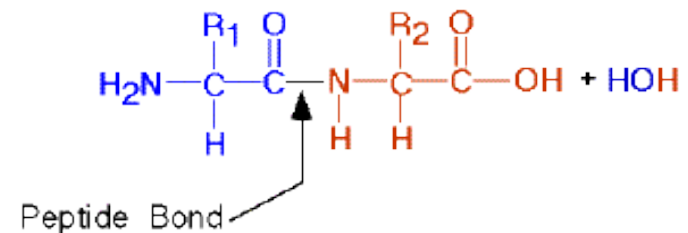
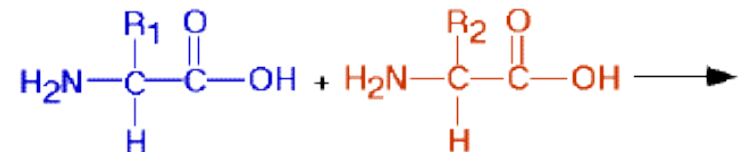
Structure of a Protein

- Concatenation of **cores**: Backbone of AA chain (a protein)
- Covalent **peptide bonds** between carboxyl and amino group (with loss of H_2O)

(A)

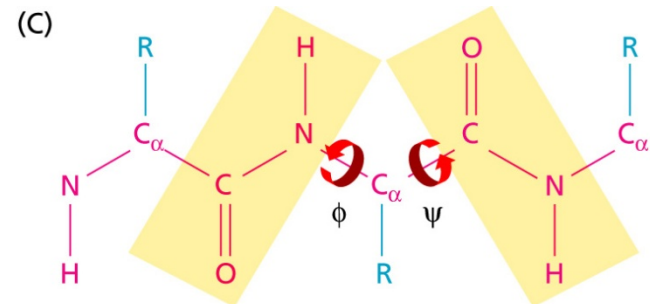
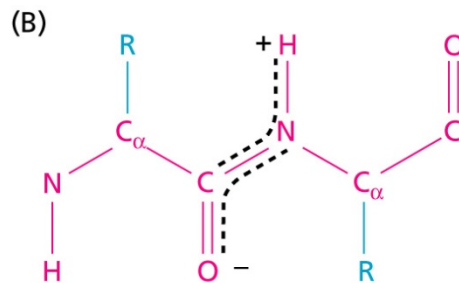
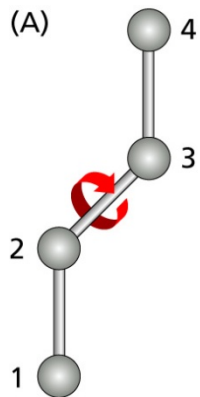


Peptide Bond Formation

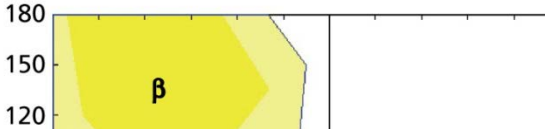


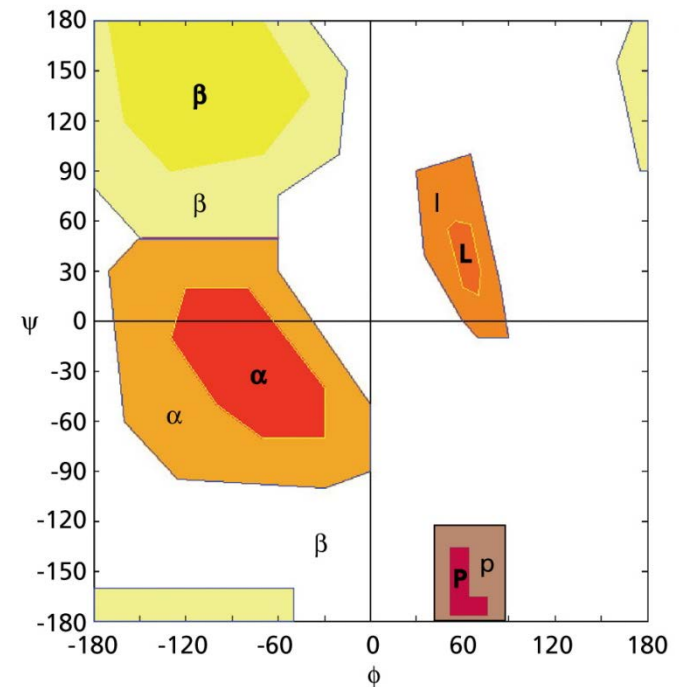
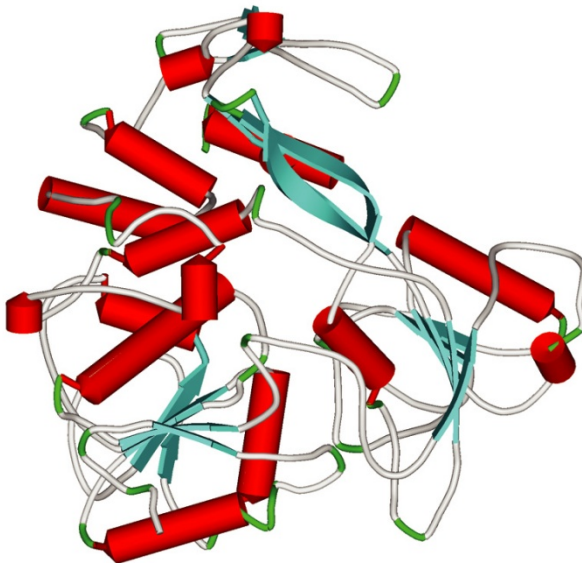
Flexibility

- In principle, every chemical bond can rotate freely
- Would allow arbitrary backbone structures
- In proteins things are more restricted
 - Peptide bond (B) is “flat” – almost no torsion possible
 - Flexibility only in the C_α -flanking bonds ϕ and ψ

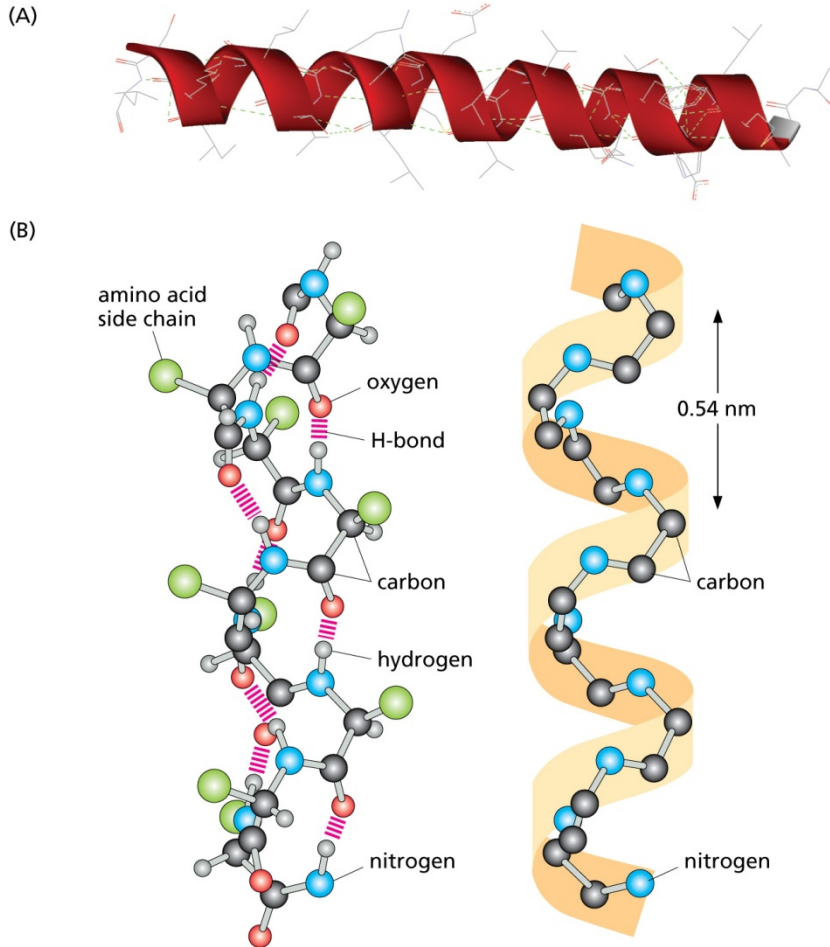


Ramachandran Plots

- Combinations of ϕ and ψ are **highly constrained**
 - Due to chemical properties of the backbone / side chains
 - Two combinations are favored: **α -helixes** and **β -sheets**
 - More detailed classifications exist
 - Angles lead to specific structures
 - **Secondary structure**
- 



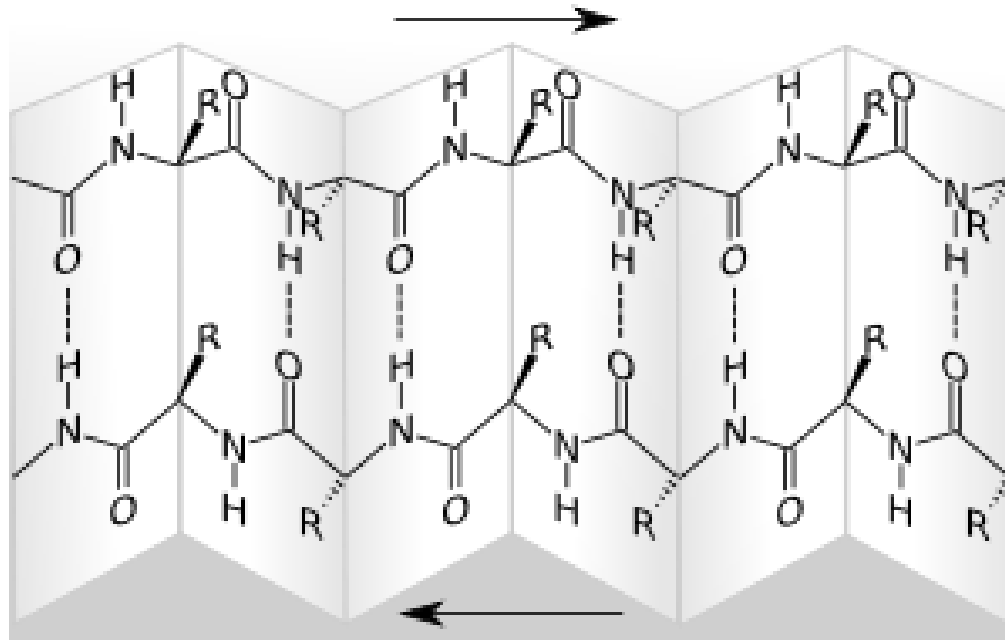
α -Helix



- Sequence of angles forming a regularly structured **helix**
- Additional bonds between amino and carboxyl groups
 - Very **stable structure**
- May have two orientations
 - Most are right-handed
- 3.4 AA per twist
- Often short, sometimes very long

β -Sheet

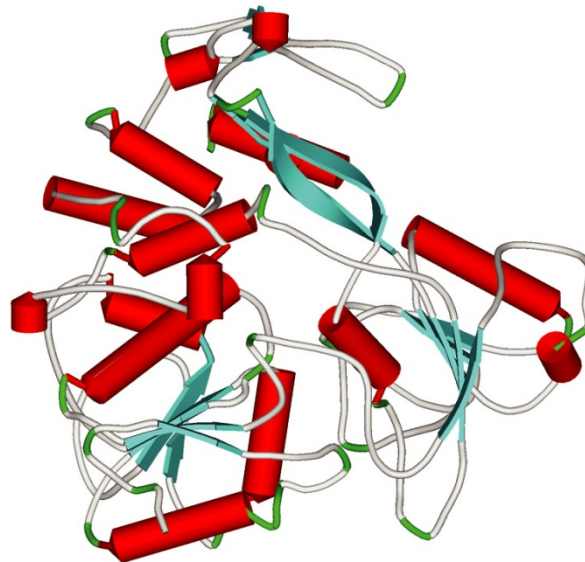
- Two linear and **parallel stretches** (β -strands)
- Strands are bound together by hydrogen bonds
- Can be parallel or anti-parallel (wrt. N/C terminus)



Quelle: Wikipedia

Other Substructures

- α -helices and β -sheets cover 50-80% of most proteins
- Other parts are called **loops or coils**
 - Usually not very important for the structure of the protein
 - But **very important for its function**
 - Often exposed on the surface; important for binding to other molecules



Importance of Secondary Structure Prediction (SSP)

- Secondary structure elements (SSE) are vital for the overall structure of a protein
- Often evolutionary well conserved
- SSE can be used to classify proteins
 - Such classes are highly correlated with function
- SSE gives important clues to protein structure
- SSP much simpler than 3D structure prediction
 - And 3D structure prediction can benefit a lot from a good SSP

Predicting Secondary Structure

- SSP: Given a protein sequence, **assign each AA** in the sequence to one of the **three classes** Helix (H), Strand (E), or Coil (-)

```
KVYGRCELAAAMKRLGLDNYRGYSLGNWVCAAKFESNFNTHATNRNTD
GSTDYGILQINSRWWCNDGRTPGSKNLCNIPCSALLSSDITASVNCAK
KIASGGNGMNAWVAWRNRCKGTDVHAWIRGCRL
```



```
KVYGRCELAAAMKRLGLDNYRGYSLGNWVCAAKFESNFNTHATNRNTD
-----HHHHHHHHHH-----EEEE-----HHHHHHHH--
GSTDYGILQINSRWWCNDGRTPGSKNLCNIPCSALLSSDITASVNCAK
----EEEEEEEEEEEEEEEEEEEE-----HHHHHH
KIASGGNGMNAWVAWRNRCKGTDVHAWIRGCRL
HHH-----EEE-----EEEE----
```

Classification

- **Classification**: Classify each AA into one of three classes
- Classification is a **fundamental problem**
 - Classify the readout of a microarray as diseased / healthy
 - Classify a subsequence of a genome as coding / non-coding
 - Classify an email as spam / no spam
- Many **different techniques**: Naïve Bayes, Regression, Decision Trees, SVMs, Neural Networks, ...
 - **Classification function** learned from properties of known objects
 - Often use same representation (feature vectors) of objects – methods exchangeable
- The following is a rather unsystematic approach
 - But simple to explain and classical for this application

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 - Chou-Fasman
 - Other methods

Chou-Fasman Algorithm

Chou & Fasman (1974). Prediction of protein conformation. Biochemistry 13

- Observation: Different AA favor different folds
 - Different AA are more or less often in H, E, C
 - Different AA are more or less often within, starting, or ending a stretch of H, E, C
- Chou-Fasman algorithm (rough idea)
 - Classifies each AA into E or H; unclassified AA are assigned C
 - Compute a score for the probability of any AA to be E (H)
 - Basis: Relative frequencies in a set of sequences with known SSE
 - In principle, assigns each AA its most frequent class
 - Add constraints like minimal length of stretches
 - Several further heuristics

Details [sketch, some heuristics omitted]

- Let $f_{j,k}$ be the relative frequency of observing AA j in class k
- Let f_k be the average over all 20 $f_{j,k}$ values
- Compute the propensity of AA j to be part of class k as
 - Relative frequency of j within class k ; probability of j given k

$$P_{j,k} = f_{j,k} / f_k$$

- Using $P_{j,k}$, classify each AA j for every class k into
 - Strong, normal, weak builder ($H_\alpha, h_\alpha, l_\alpha$)
 - Strong, weak breaker (B_α, b_α)
 - Indifferent (i_α)

Concrete Values

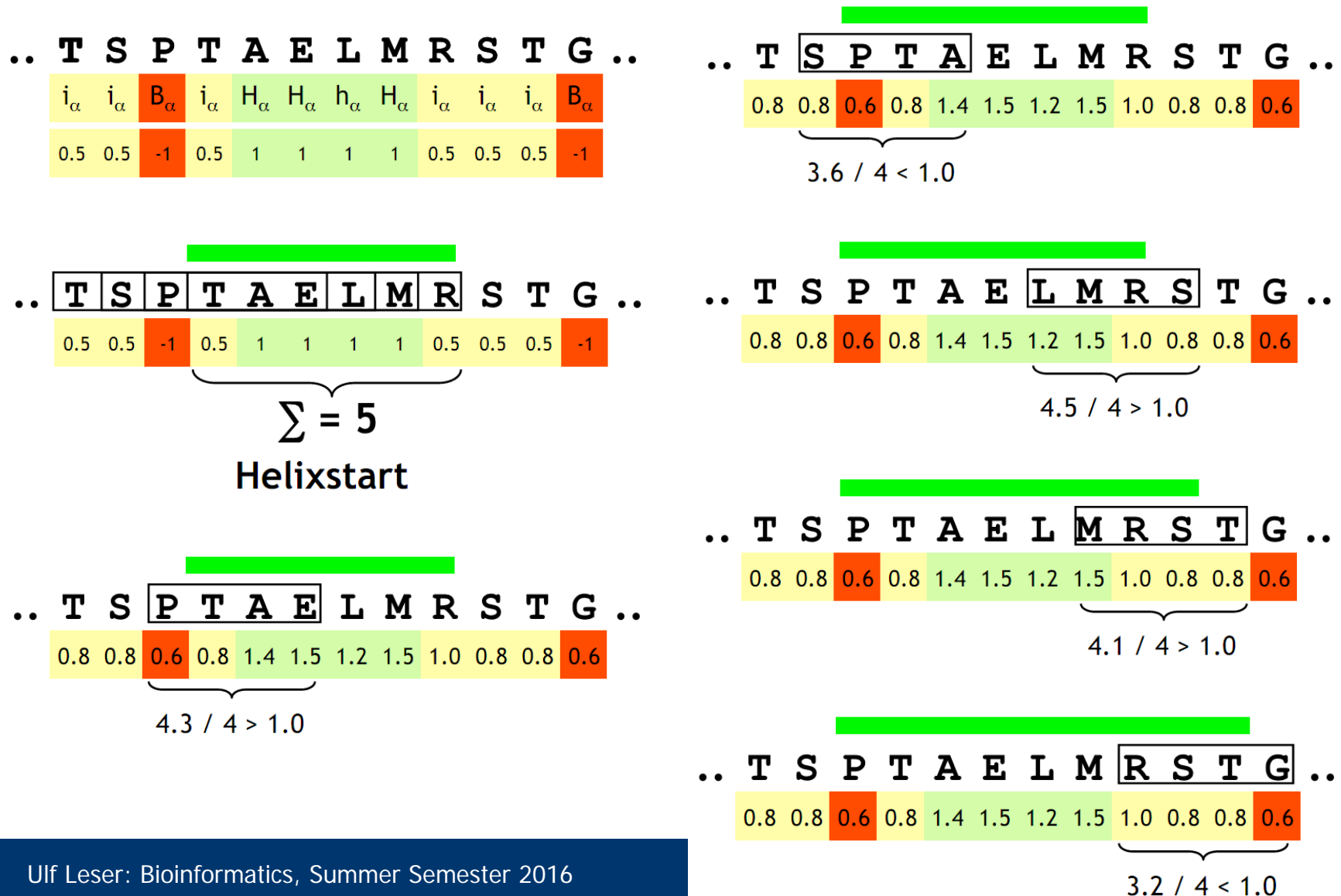
- Originally computed on only 15 proteins (1974)

AS	P_α	Klasse	AS	P_β	Klasse	AS	P_α	Klasse	AS	P_β	Klasse
Glu	.53	H_α	Met	.67	H_β	Ile	1.00	I_α	Ala	0.93	I_β
Ala	1.45		Val	1.65		Asp	0.98	i_α	Arg	0.90	i_β
Leu	1.34		Ile	1.60		Thr	0.82		Gly	0.81	
His	1.24	h_α	Cys	1.30	h_β	Ser	0.79		Asp	0.80	
Met	.20		Tyr	1.29		Arg	0.79		Lys	0.74	b_β
Gln	1.17		Phe	1.28		Cys	0.77		Ser	0.72	
Trp	1.14		Gln	1.23		Asn	0.73	b_α	His	0.71	
Val	1.14		Leu	1.22		Tyr	0.61		Asn	0.65	
Phe	1.12		Thr	1.20		Pro	0.59	B_α	Pro	0.62	
Lys	1.07	I_α	Trp	1.19		Gly	0.53		Glu	0.26	B_β

Algorithm for Helices

- Go through the protein sequence
- Score each AA with 1 (H_{α}, h_{α}), 0.5 (I_{α}, i_{α}), or -1 (B_{α}, b_{α})
- Find helix cores: subsequences of length 6 with an aggregated AA score ≥ 4
- Starting from the middle of each core, shift a window of length 4 to the left (then to the right)
 - Compute aggregated score A using values $P_{j,k}$ inside the window
 - If $A \geq 4$, continue; otherwise stop
- Similar method for strands
- Conflicts (regions assigned both H and E) are resolved based on aggregated scores

Example [Source: O. Kohlbacher, "Strukturvorhersage"]



Performance

- Accuracy app. 50-60%
 - Measured on per-AA correctness
- Prediction is **more accurate in helices** than in strands
 - Because helices build **local bridges** (hydrogen bounds between the turns; each AA binds to the +4 AA)
- General problem
 - Secondary structure is not only a local problem
 - Looking **only at single AAs** is not enough
 - Note: Scores are based on individual AA; aggregation by summation assumes **statistical independence** of pairs, triples ... in a class
- One needs to include the **context of an AA**

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Classes of Methods

- First generation: Properties of single AA only
 - Accuracy: 50-60%
 - E.g. Chou-Fasman (1974)
- Second generation: Include info. about neighborhood
 - Accuracy: ~65%
 - E.g. GOR (1974 – 1987)
- Third generation: Include info. from homologous seq's
 - Accuracy: ~70-75%
 - E.g. PHD (1994)
- Forth generation: Build ensembles of good methods
 - Accuracy: ~80%
 - E.g. Jpred (1998)

Further Reading

- Gerhard Steger (2003). "Bioinformatik – Methoden zur Vorhersage von RNA- und Proteinstrukturen", Birkhäuser, chapter 8,10,11,13
- Zvelebil, M. and Baum, J. O. (2008). "Understanding Bioinformatics", Garland Science, Taylor & Francis Group, chapter 2, 11, 12 (partly)