

Computergestützte Strukturbiologie (Strukturelle Bioinformatik)

Membrane proteins

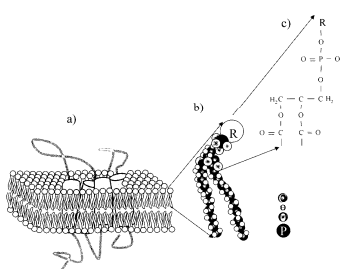
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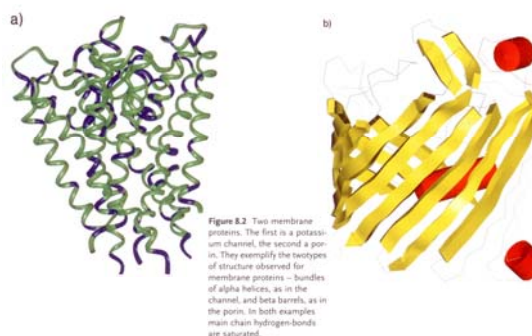
Membrane proteins

- About 25% of all human proteins are membrane proteins.
- The majority of proteins that are drug targets are membrane proteins.
- However, less than 1% of the protein structures in the PDB are membrane proteins.
- Membrane proteins are difficult to crystallize.
- Membrane proteins are insoluble in water and thus difficult to study by NMR spectroscopy.

Membrane with an embedded protein



Two membrane proteins



Membrane protein structure and function

- Membrane proteins are either α -helix bundles or β -barrels
- Most of the important membrane proteins are α -helical. β -barrel membrane proteins are found only in the outer membrane.
- Functions:
 - Transport of ions/molecules through the membrane
 - Signal transduction (e.g. G-protein coupled receptors; GPCR)

Hydrogen bonding in membrane proteins

- In soluble proteins hydrogen bonds can be formed between protein atoms and to the solvent (water).
- In the hydrophobic environment of membrane proteins only intra-protein hydrogen bonds can be formed.
- The energy cost of burying a pair of unsatisfied hydrogen bond donors/acceptors is high: ~ 4 kcal/mol.
- Transmembrane segments that are α -helices or β -strands involve all main chain polar atoms in hydrogen bonds.
- Random coil structures within the membrane bilayer is energetically unfavorable and rare.

Membrane protein structure prediction

- The present data base of known membrane protein structures is small.
- Less structural variability than for soluble proteins
- Prediction of transmembrane helices based on secondary structure prediction and hydrophobicity

Which features of membrane protein structure can be predicted?

- Discrimination between globular and membrane proteins
- Identification of individual transmembrane helices
- Location of loops on the inside/outside of the membrane
- Topology

Chou-Fasman propensities for α , β , and coil secondary structure

Table 7.1 The Chou and Fasman parameters for secondary structure prediction.

Amino acid	P_{α}	P_{β}	P_C	Amino acid	P_{α}	P_{β}	P_C
Alanine	1.42	0.83	0.66	Leucine	1.21	1.30	0.59
Arginine	0.98	0.93	0.95	Lysine	1.14	0.74	1.01
Aspartic acid	1.01	0.54	1.46	Methionine	1.45	1.05	0.60
Asparagine	0.67	0.89	1.56	Phenylalanine	1.13	1.38	0.60
Cysteine	0.70	1.19	1.19	Proline	0.57	0.55	1.52
Glutamic acid	1.39	1.17	0.74	Serine	0.77	0.75	1.43
Glutamine	1.11	1.10	0.98	Threonine	0.83	1.19	0.96
Glycine	0.57	0.75	1.56	Tryptophan	1.08	1.37	0.96
Histidine	1.00	0.87	0.95	Tyrosine	0.69	1.47	1.14
Isoleucine	1.08	1.60	0.47	Valine	1.06	1.70	0.50

Chou-Fasman secondary structure prediction

	F	T	L	E	W	F	L	S	H	C	H	I	H	K	Y
P_{α}	0.57	0.83	1.21	1.39	1.08	1.13	1.21	0.77	1.00	0.70	1.00	1.08	1.00	1.14	0.69
P_{β}	0.55	1.19	1.3	1.17	1.37	1.38	1.30	0.75	0.87	1.19	0.87	1.6	0.87	0.74	1.47
Predicted	E	E	E	E	E	E	E	E	E	E	E	E	E	E	E
Observed	H	H	H	H	H	H	H	H	H	H	H	H	H	H	H

	F	S	K	S	T	L	I	H	Q	G	E	K	A	E	T
P_{α}	0.57	0.77	1.14	0.77	0.83	1.21	1.08	1.00	1.11	0.57	1.39	1.14	1.42	1.39	0.83
P_{β}	0.55	0.75	0.74	0.74	1.09	1.20	1.60	0.87	1.18	0.78	1.17	0.74	0.83	1.17	1.09
Predicted	E	E	E	E	E	E	E	E	E	E	E	E	E	E	E
Observed	E	E	E	E	E	E	E	E	E	E	E	E	E	E	E

	I	Y	Y	I	V	K	G	S	V	A	V	L
P_{α}	1.21	0.69	0.69	1.08	1.06	1.14	0.57	0.77	1.06	1.42	1.06	1.21
P_{β}	1.30	1.47	1.47	1.60	1.70	0.74	0.75	0.75	1.70	0.83	1.70	1.30
Predicted	E	E	E	E	E	E	E	E	E	E	E	E
Observed	E	E	E	E	E	E	E	E	E	E	E	E

Figure 7.3 An example of the application of the Chou and Fasman method to a protein. It is apparent that the predicted and experimental structures only have limited overlap.

Neural network nodes

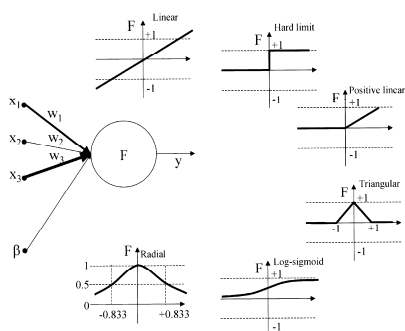


Figure 7.5 A node of a neural network is a computational unit that transforms input values x_i into an output value y . The input values can have different weights and the transformation can be different. A few examples of "transfer functions", used to compute the output of the node as a function of the weighted sum of its inputs, are shown in the figure.

Neural network layers

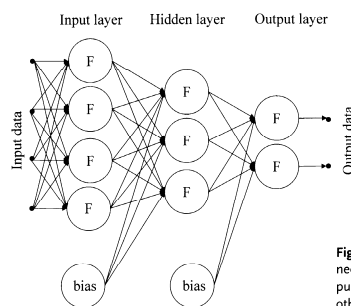
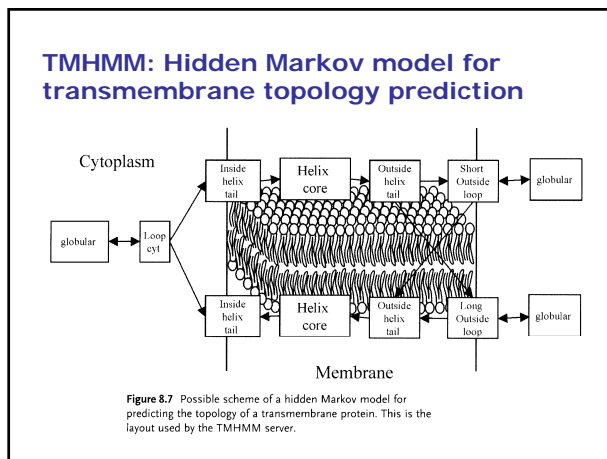
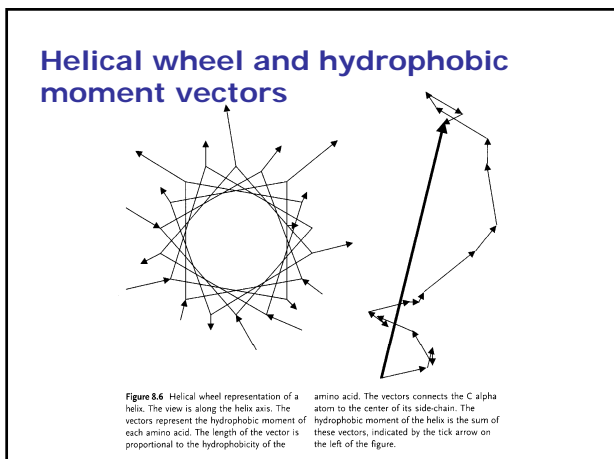
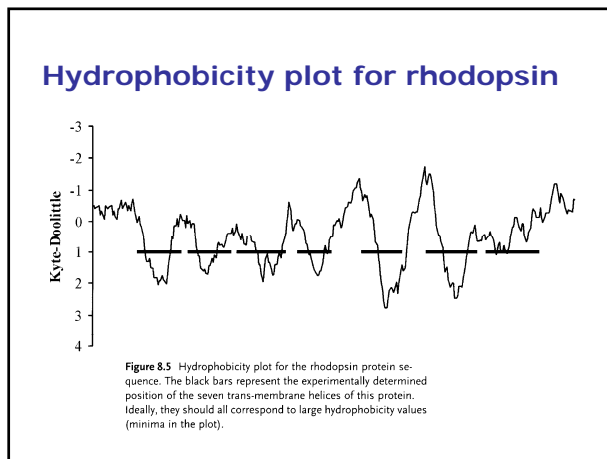
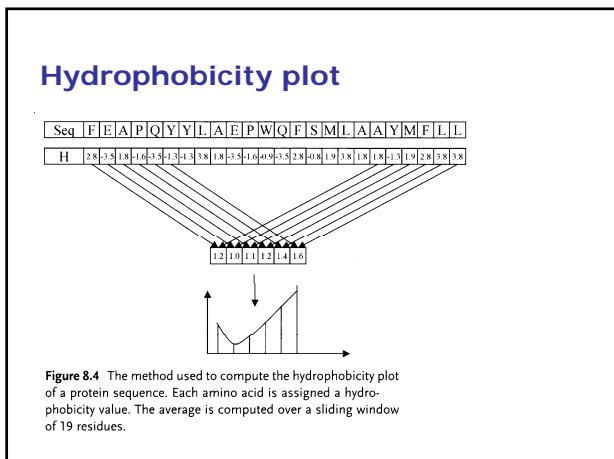
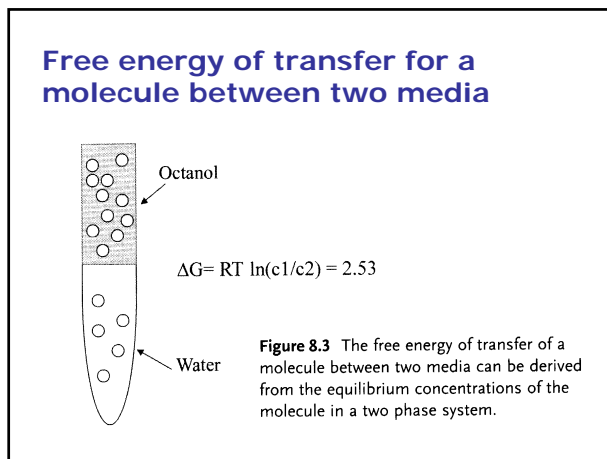
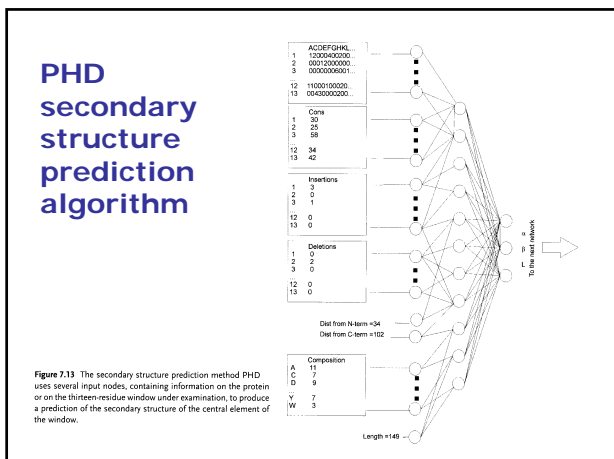


Figure 7.6 Several nodes can be connected in a network such that the output of some nodes is the input of others.



TMHMM performance

- **Discrimination globular/membrane: sensitivity and specificity > 98%**
- **Correct topology: 65-70%**
- **Single transmembrane helix identification: sensitivity 96%, specificity 98%**
- **Training set: 160 membrane proteins, 650 globular proteins**

<http://www.cbs.dtu.dk/services/TMHMM/>
Krogh et al., *J. Mol. Biol.* 305, 567-580 (2001)

Literatur

- Anna Tramontano: *Protein Structure Prediction*, Wiley-VCH, 2006.
- A. Krogh, B. Larsson, G. Von Heijne, E. L. L. Sonnhammer. Predicting transmembrane protein topology with a hidden markov model: application to complete genomes. *J. Mol. Biol.* 305, 567-580 (2001)