

Metody ograniczenia przestrzeni poszukiwań w modelowaniu nieznanych struktur białkowych

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Motywacja biologiczna

CEL: Przewidywanie struktury przestrzennej białek wyłącznie z sekwencji aminokwasów.

OBIEKT: Białka o znanej sekwencji i nieznanej strukturze

Jak wielu białek dotyczy ten problem ?

UniProtKB/Swiss-Prot – 0,5 mln sekwencji białkowych

UniProtKB/TrEMBL - 20 mln sekwencji białkowych przetłumaczonych z DNA

PDB – 90 tys. trójwymiarowych struktur białkowych

Źródło: UniProt i PDBe

Klasyczne metody modelowania struktury trzeciorzędowej

- Ab-initio / de Novo
- Porównawcze:
 - Homologiczne
 - Nawlekanie
- Statystyczne (uczenie maszynowe)

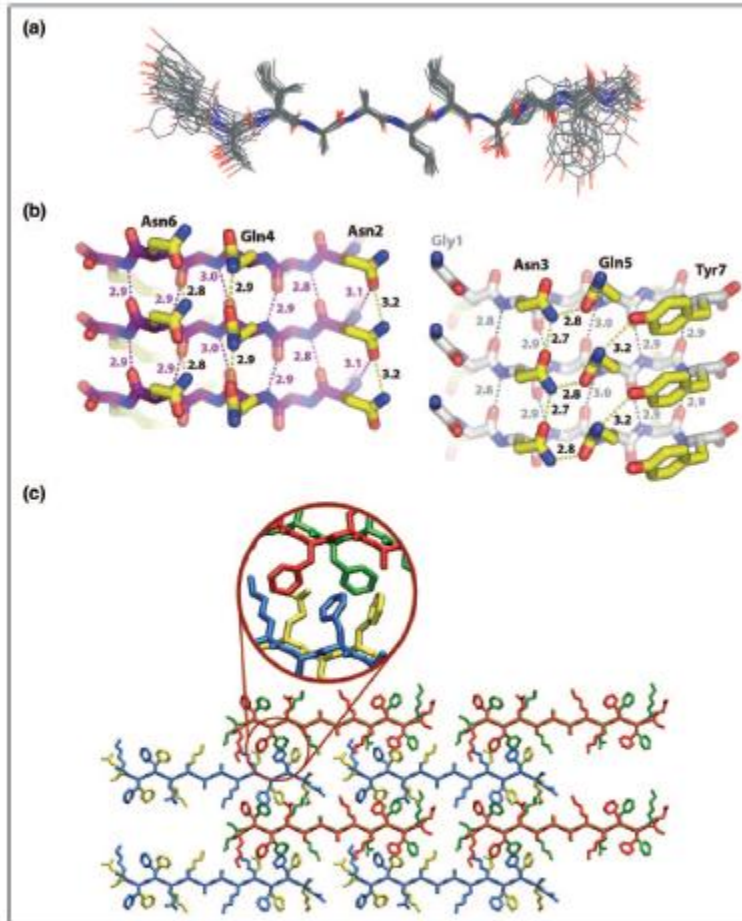
Nasze kierunki badań

- Ocena jakości modeli w oparciu o relację **struktura-funkcja**
 - charakterystyka przepływowa kanałów jonowych
 - zastosowanie **ontologii genów** - GO
- Zastosowanie miejsc kontaktowych białek do modelowania ich struktury – **uczenie maszynowe. Ontologia miejsc kontaktowych.**
- **Amyloidy** - różne struktury białek o tej samej sekwencji. **Uczenie maszynowe a modelowanie fizykochemiczne**

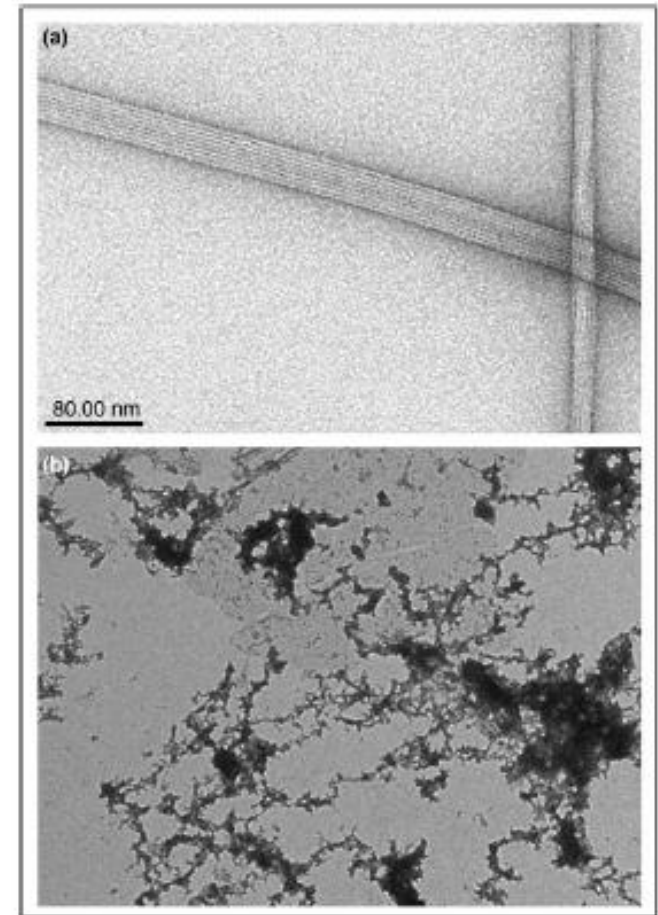
**How proteins forming
amyloids can be recognized
computationally**

Amyloid structure

„steric zippers”



High-resolution structures of peptides in amyloid fibrils. (a) Structure derived from magic angle spinning NMR of fibrils grown from a peptide fragment of transthyretin [22**]. (b) Crystallographic model of a fibril comprising a peptide sequence derived from the yeast prion Sup35 [26]. (c) High-resolution model from crystallography and electron microscopy of fibrils of a synthetic peptide sequence [25**].



Electron micrographs of the macromolecular morphology of protein aggregates and amyloid fibrils. (a) Amorphous aggregates (of the fusion proteins between the NusA protein and the Alzheimer's β -peptide) and (b) amyloid fibers (of the hexapeptide STVIIIE [23]). Both aggregates have a high content of β -structure, but dramatically different morphologies when seen through an electron microscope.

Exemplary diseases

- Alzheimer disease (amylo β -42, τ)
- Parkinson disease (α -synuclein)
- Diabetes type II (islet amyloid polypeptide - IAPP, amylin)
- Amyotrophic Lateral Sclerosis ALS (SOD)
- Huntington disease (huntington)
- Creutzfeldt-Jacobs (prion, e.g. sup-35)
-

Hot-spots of hexapeptides

Table 1

Amyloid fibril-forming proteins and amyloidogenic determinants experimentally investigated [10 and references therein], and predicted according to [10,14,8]

Amyloidogenic protein	Experimentally determined amyloidogenic regions [10]	Predicted high packing density peptides [10] ^a	Conformational switches according to SecStr [14] ^a (numbering)	Conformational switches according to SecStr [14] (peptide sequence)	Pattern proposed in [8] according to PATTINPROT [22]
Prion, 253, 22 P04156	132–160	136–141 147–152			
	178–193	<u>174–178</u> 180–184	175–183*	FVHDCVNIT	DCVNIT
		<u>211–217</u> 232–236	209–215*	VVEQMCI	EQMCIT
		<u>240–250</u>	242–251*	LLISFLIFLI	LIFLIV
Amyloid- β , 770, 17 672–713 = > A β 42 672–711 = > A β 40 P05067	686–695 702–710	<u>687–692</u> <u>703–708</u>	686–691* 718–721	QKLvFF ITLV	QkLvFF, KLVFFA ITLvml
Cystatin-C, 146, 26 P01034		<u>73–77</u> <u>87–92</u> <u>121–131</u>	73–77* 89–94* 120–123*	LQVVR FLDVEL KAFC	LQVvrA NYFLdV KAFcsF
Amylin, 89, 22 P10997	36–41 42–49	<u>46–51</u>	44–49* 74–79	RLANFL NAVEVL	RLaNFL NAVEvL
Calcitonin, 141, 25 P01258	40–44	<u>48–54</u> 90–95	49–59*	LLLAALVQDYV	LVQdYV
Atrial natriuretic factor, 153, 25 P01160		<u>116–120</u>	37–44 116–119*	LMDFKNLL RALL	MDFknL RALLtA

Why modeling?

Non-informed experimental tests
(variations with repetition)

$$20^6 = 64\,000\,000$$

Methods in modeling

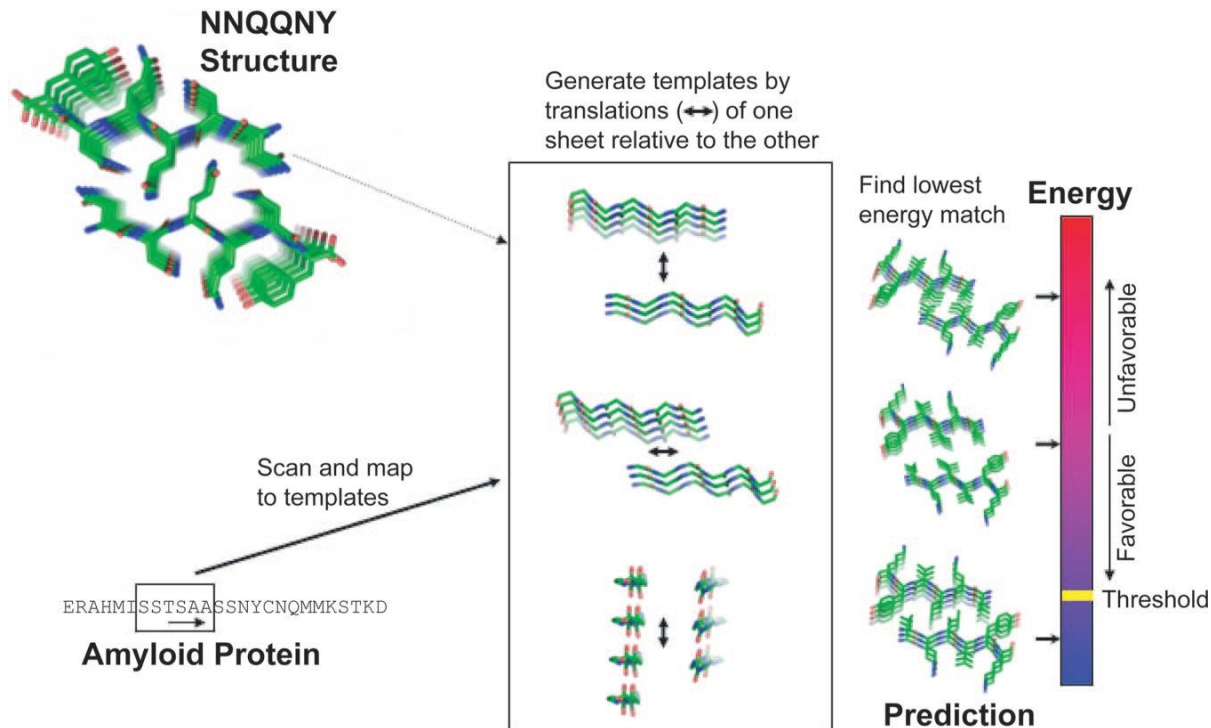
Physico-chemical methods in modeling

- Concept of β - aggregates
- Packing density (contact sites) – **FoldAmyloid** (Galzitskaya OV, Garbuzynskiy SO, Lobanov MY ,*PLoS Comp.Biol.* 2006)
- Concept of α - β switch (conversion from α -helix to β -sheet)
- Stability of supersecondary structure eg. **Profile 3D**

Statistical methods

- **Waltz** (Maurer-Stroh 2010) – frequency characteristics
- Bayess, Decission Tree (David 2010), bacterial strands
- **Pafig** (Tian 2009), SVM, machine learning based on AAindex features (**Erroneous dataset!**)

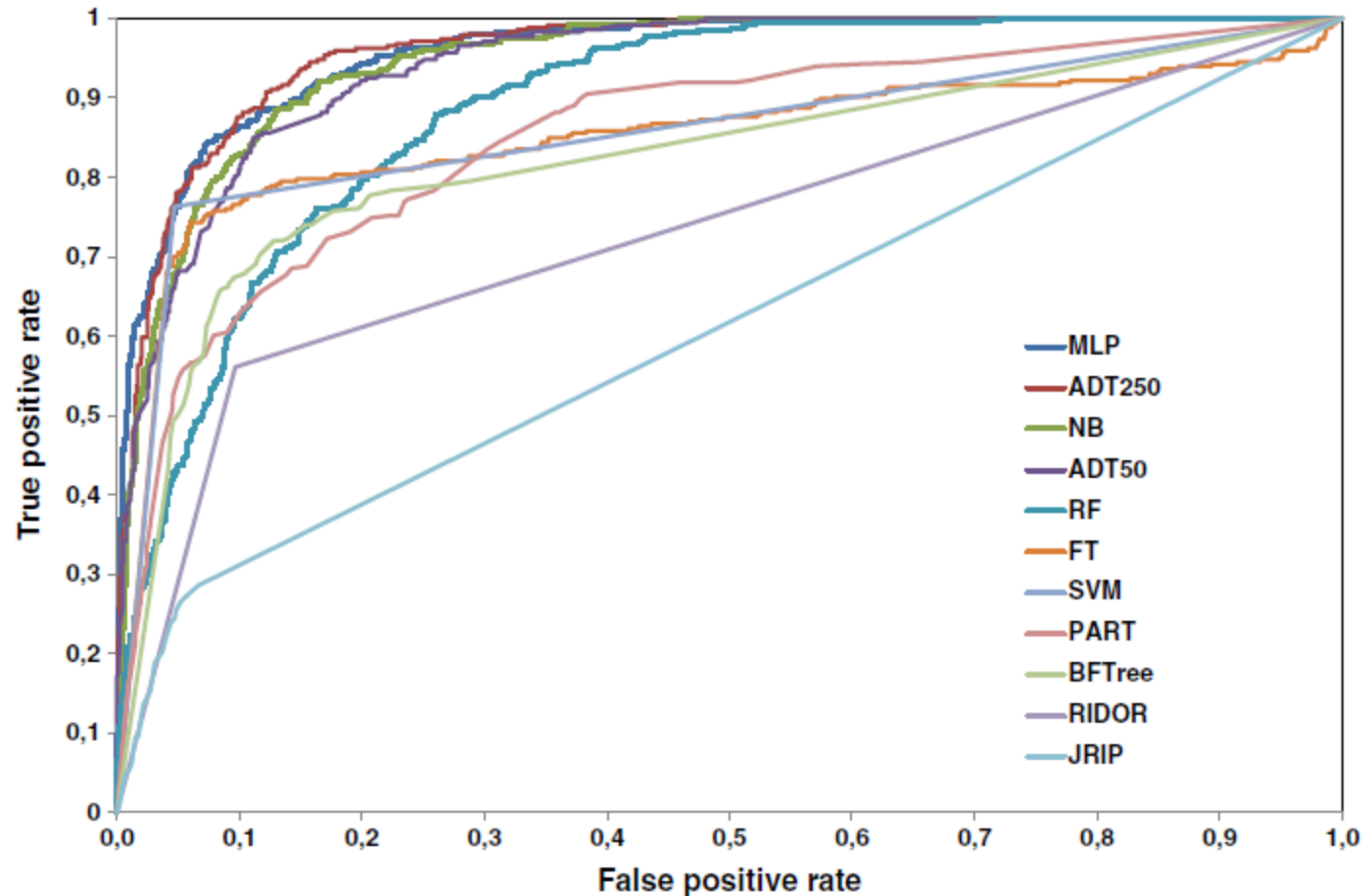
Machine learning methods for hexapeptides



Simplified **3D profile** method gives 93.5% of identity with the original method of **ZipperDB**

Thompson et al. PNAS 2006

Machine learning methods



Classification efficiency

- **18–20 CPU-hours** full profile 3D / 1 hexapeptide
- **0.5 CPU-hours** simplified profile 3D
- **Seconds** machine learning

J. Stanislawski, M. Kotulska, O. Unold, Machine learning methods can replace 3D profile method in classification of amyloidogenic hexapeptides, **BMC Bioinformatics** 2013

Recognition of hot-spots based on
site specific aminoacid pairwise
co-occurrence

Recognition of hot-spots based on site specific aminoacid pairwise co-occurrence

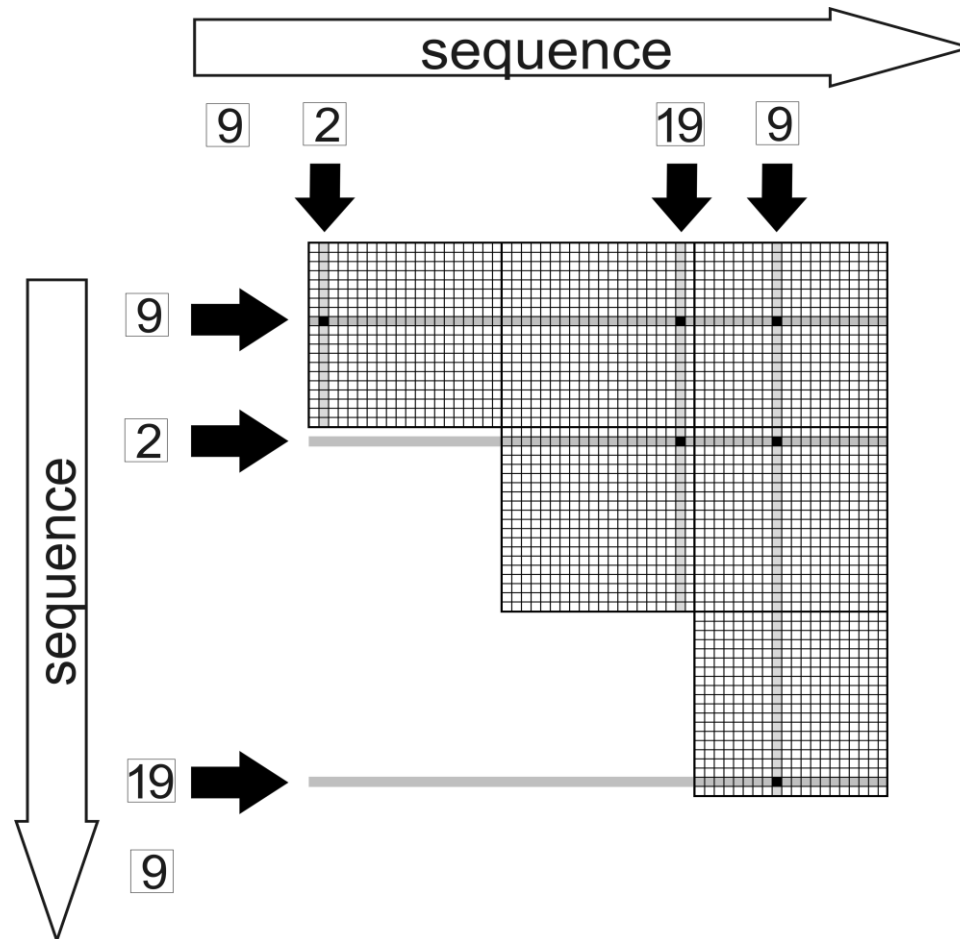


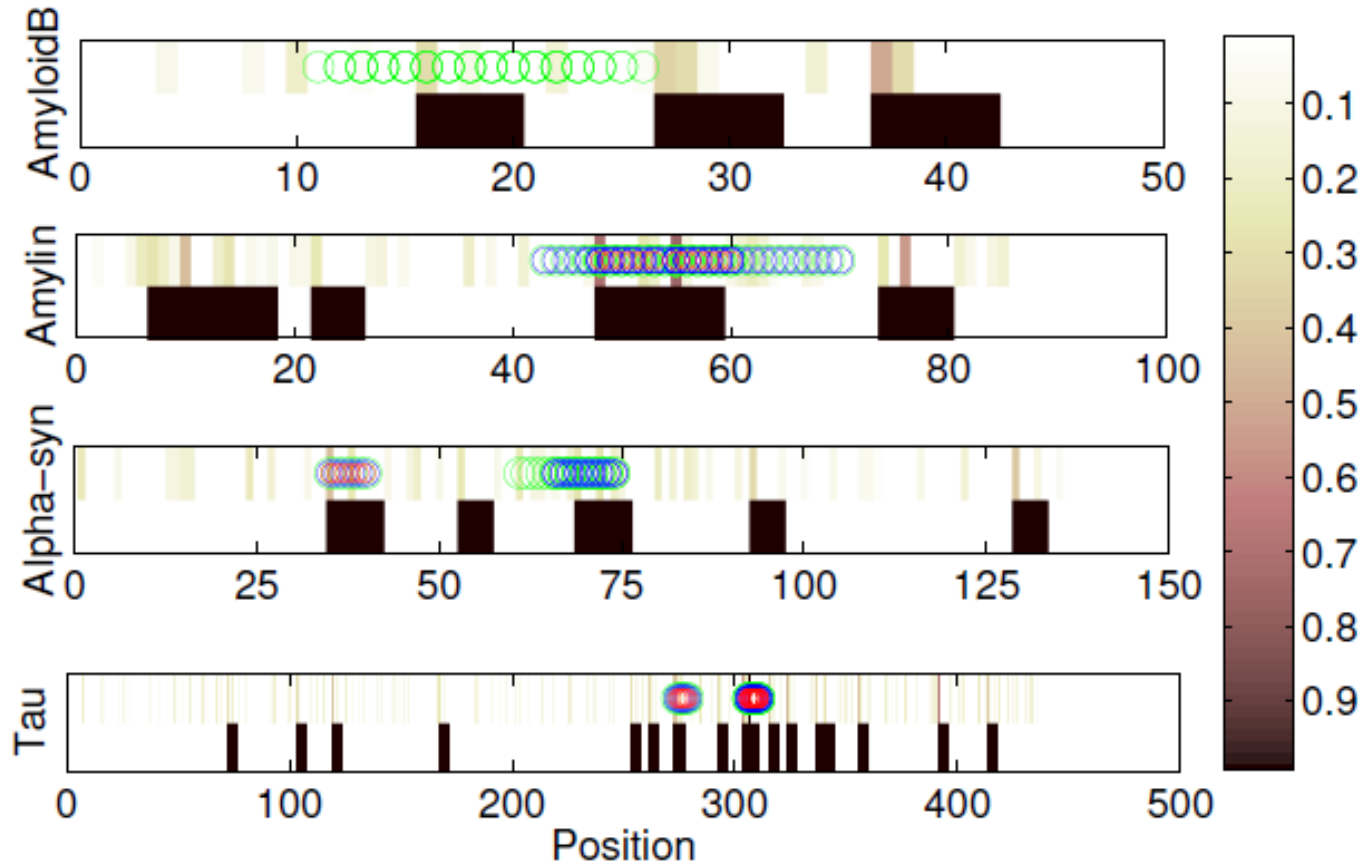
TABLE 2.

		2	3	4	5
TG	1-	F-F, F-T	F-A, I-L	F-E, Y-Y, I-I	F-D, Y-Y,
	2-		F-A, F-L	F-E, L-Y, V-I	F-D, V-V
	3-			A-E, L-Y, L-I	A-D, Q-V, T-V
	4-				E-D, I-V
AmylHex		L-T, S-L	V-A, F-V, N-V, Q-V	L-I	S-I, V-A,
	2-		I-V, E-A, L-V	E-L	E-A, F-I
	3-			A-L, V-T, V-I	V-I
	4-				I-I, L-A
Waltz	1-	S-F, Q-F, G-V, Y-V	Y-N, Q-I	Q-I, S-I, I-I, F-I, G-	G-F, F-Q, Q-S
	2-		V-L, V-N, F-L, F-N,	F-I, V-Y, V-F, V-I	V-F, V-V, V-I
	3-			L-I, N-Y, E-Y	L-V, N-F, L-F
	4-				I-V, I-M, I-F, F-S
All + sup35	1-	S-F, G-V	S-V, Y-N	S-I, Y-Y, L-Y, I-I	S-I, Y-Q
	2-		F-L, T-V	V-I, T-I, V-Y, F-I	R-R, T-I, V-I
	3-			V-I, L-I, N-Y	V-I, L-V
	4-				I-I, I-F, Y-R, Y-I
3D profile	1-	A-I, T-V	S-V, T-I	S-I, V-I, T-I	S-I, V-F
	2-		T-V, V-I	T-I, V-I, L-I	T-I
	3-			V-I, I-I, V-N	V-I, L-I, V-V
	4-				I-I, T-C

Correlated pairs of aminoacids

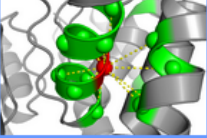
The result of machine learning on each of 4 training datasets and their combination with the window of length 5. The correlated pairs of aminoacids, increasing the chance of amyloidogenicity, are shown with their most probable locations.

Finding hot spots in full length proteins



Black blocks location of amyloidogenic segments with $w_i = 0.14$, (Sp=60% on Waltz set). The brown - different w_i values assumed. The circles experimentally by different groups, working on protein fragments of various lengths (green – above 16, blue -11, red - 7).

FISH Amyloid



General information

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The main parameters

Threshold

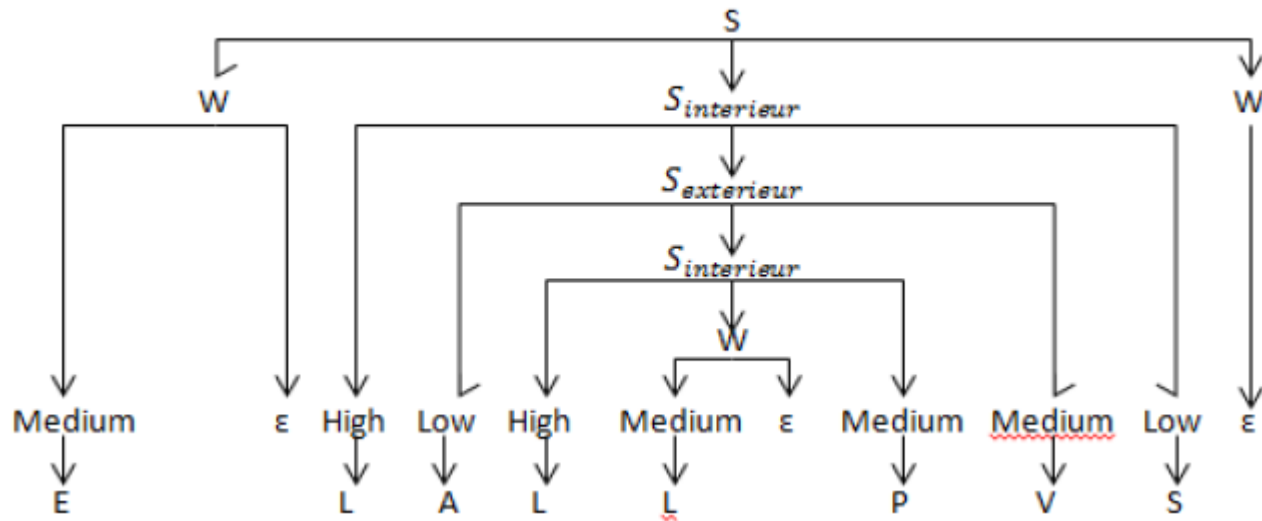
Prediction

Target sequence in one letter code

Description can be found [here](#)

Gasior P, Kotulska M, FISH Amyloid – a new method for finding amyloidogenic segments in proteins based on site specific co-occurrence of aminoacids, BMC Bioinformatics 2014, in print

Exemplary grammar (hydrophobicity)



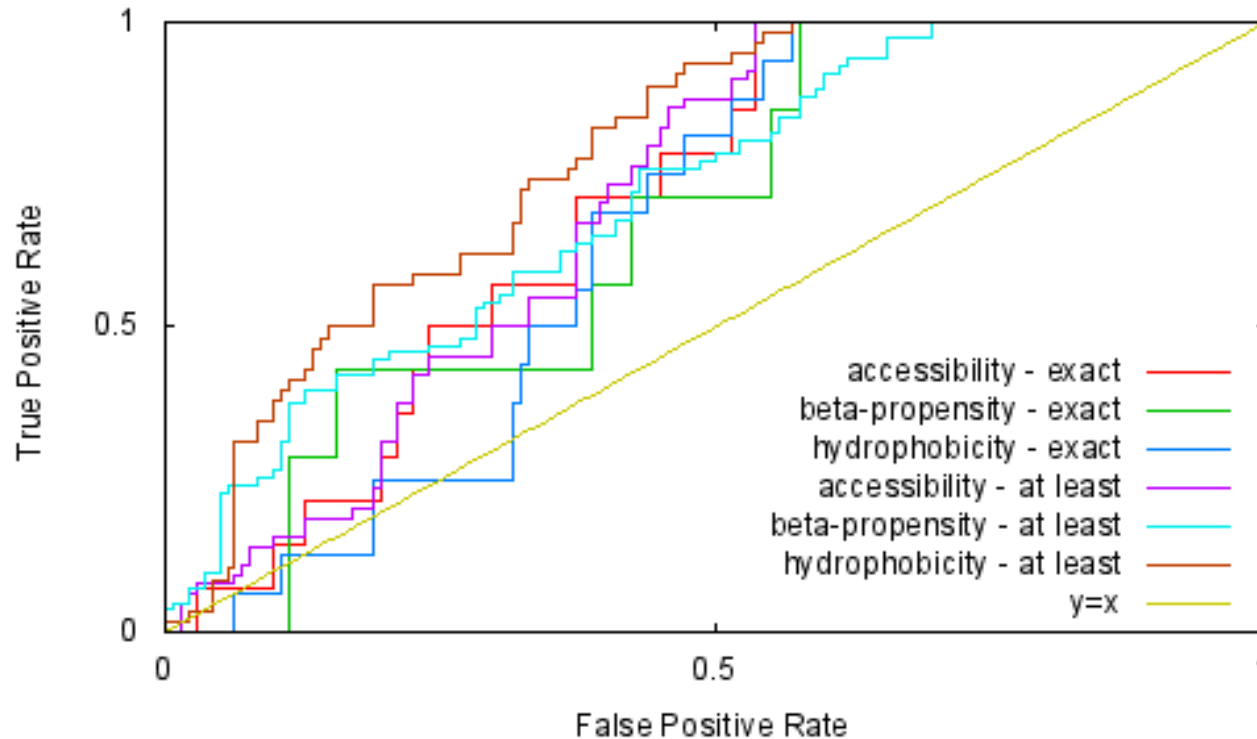
True dependencies according to DSSP



Dependencies found by the grammar

Recognition of double strands with PCFGs

The grammar designed for β -barrels



ROC curves of grammars trained on Sawaya dataset (experimental double strands) and tested on double Waltz and double Waltz negative

Grammar scheme for barrels: Waldispühl J, Berger B, Clote P, Steyaert JM (2006): Predicting Transmembrane beta-barrels and interstrand residue interactions from sequence. *Proteins*, 65, 61-74.

SUMMARY

- Amyloidogenicity may be attributed to short segments in sequences, called hot-spots
- Very limited experimental data
- Hot spots can be recognized with various computational methods
- Transmembrane channels, probably formed by amyloidogenic segments can be a modeling goal

Publikacje

Stanislawski Jerzy, Kotulska Małgorzata, Unold Olgierd: Machine learning methods can replace 3D profile method in classification of amyloidogenic hexapeptides / Jerzy Stanislawski, Malgorzata Kotulska, Olgierd Unold. **BMC Bioinformatics** [Dokument elektroniczny]. 2013, vol. 14, [art.] 21,

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Dyrka Witold, Bartuzel Maciej, Kotulska Małgorzata, Optimization of 3D Poisson-Nernst-Planck model for fast evaluation of diverse protein channels, **PROTEINS: Structure, Function, and Bioinformatics.** 81(10):1802-22, 2013

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Kurczyńska Monika, Cichowski Wojciech: Ion transport simulation of transmembrane protein, Biocybernetyka i inżynieria biomedyczna [Dokument elektroniczny]: XVIII krajowa konferencja naukowa, 10-12 października 2013 - Gdańsk / red. Adam Bujnowicz, Jerzy Wtorek. [Gdańsk : Wydział Elektroniki, Telekomunikacji i Informatyki Politechniki Gdańskiej, 2013]. s. 1-9

Monika Kurczyńska: Diffusion coefficient and eletrostatic potential in brownian dynamics simulations, [W:] 11th Students' Science Conference, 03-06 październik 2013 - Będlewo. Wrocław: Oficyna Wydawnicza Politechniki Wrocławskiej, 2013. s. 475-480.

Cichowski Wojciech, Kurczyńska Monika: Ion Current Analyser (ICA) - narzędzie do analizy wyników symulacji transportu jonów przez kanały jonowe, Otwarta Innowacja. 2013, nr 1/2, s. 16-25.

Konopka Bogumił M.: A procedure for automated contact map-based reconstruction of protein structural models, Postępy Inżynierii Biomedycznej, pod red. Lucyna Leniowska, Zbigniew Nawrat, Rzeszów, Uniwersytet Rzeszowski, 2013. s. 63-73, ISBN 978-83-63151-02-7

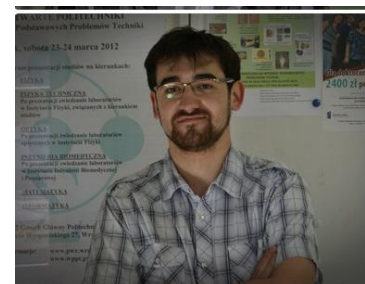
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Probabilistic Context Free Grammars (PCFG)

Only physicochemical properties of residues

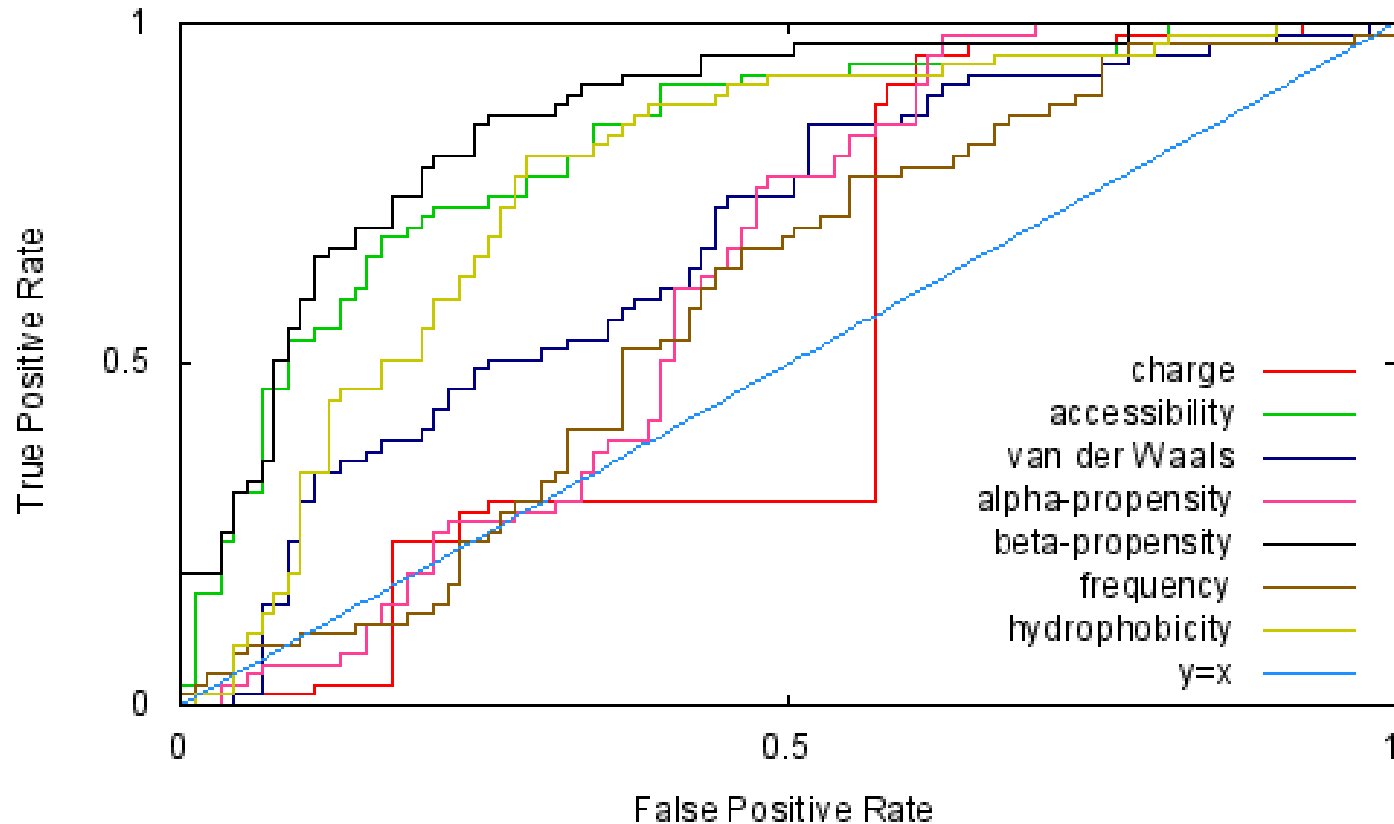
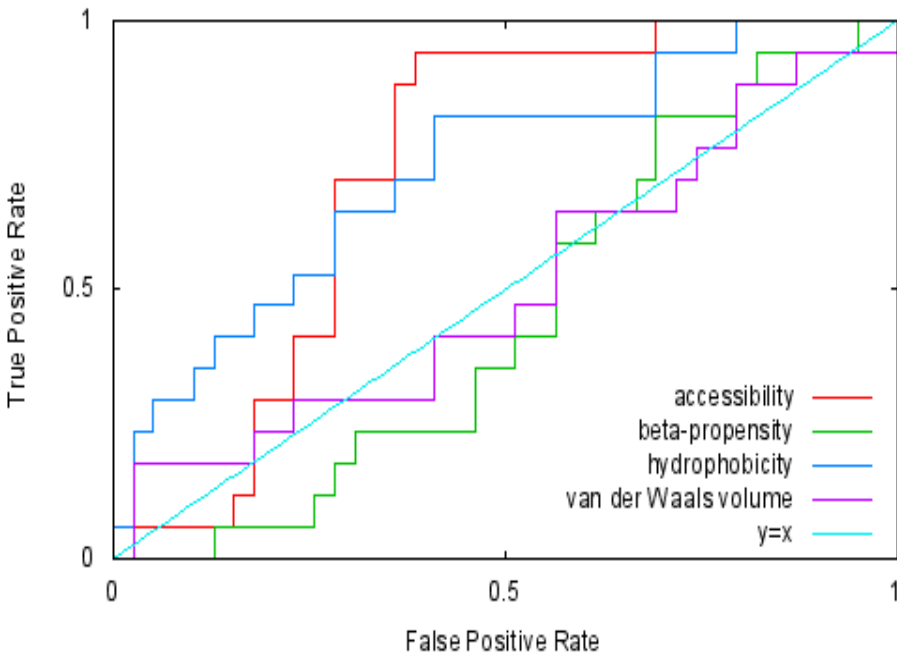
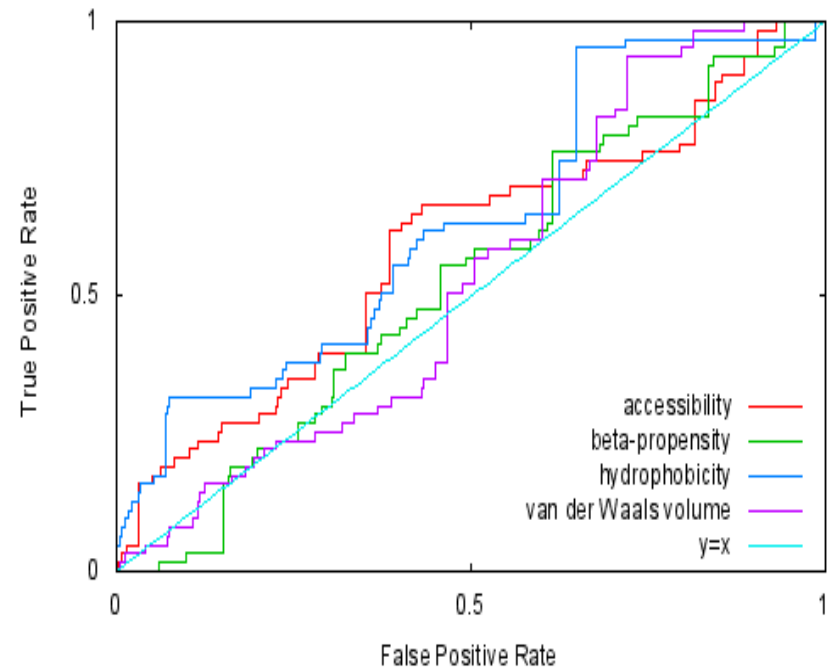


Figure 5.1.1: ROC curves of grammars trained on **Waltz positive** dataset and tested on **AmylHex** datasets.

Beta-aggregates versus amyloids



ROC curves of the grammars trained on Waltz and tested on Tango datasets (beta-aggregates versus amyloids).



ROC curves of the grammars trained on Waltz and tested on Tango datasets (beta-aggregates versus amyloids), when using a **scanning window of 6 residues**