N-gram analysis of amyloid data

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Amyloids are short proteins associated with a number of clinical disorders, for example Alzheimer's or Creutzfeldt-Jakob's diseases. Despite their variability in size, amino acid composition, most amyloidogenic sequences form cytotoxic aggregates, although there are some biologically functional [BU15]. The hallmark trait of amyloids is the presence of characteristic short sequences of amino acids, called hot-spots, amyloids can create zipper-like β -structures [F12]. Although studies investigating properties of amyloidogenic sequences have already been conducted, the newly established AmyLoad database facilities large-scale analysis of amyloids [WK15].

Among commonly acclaimed methods of predicting amyloids, FISH Amyloid [GK14] focuses more on the putative motifs of hot spots. To expand its model by considering longer and more complicated motifs, we used n-gram analysis. N-grams (k-mers) are vectors of n characters derived from input sequences. The number of possible n-grams is equal to n^u , where u is the length of the alphabet (4 in case of nucleic acids and 20 in case of proteins). To deal with the dimensionality of the problem, we implemented QuiPT (Quick Permutation Test) in biogram software [BSL15], which performs an exact test instead of a large number of permutations.

To reduce the dimension even more, we grouped amino acids into clusters based on their physicochemical properties potentially important in the amyloid type of aggregation. The features include several scales quantitatively representing hydrophobicity, size and accessibility derived from AAIndex data base as well as propensity to form contact sites derived in [WK14].

The n-gram model, trained on the data from AmyLoad database, is validated through amyloid prediction framework using random forests. The preliminary analysis of the amyloidogenic sequences not only can facilitate prediction of amyloidsyield but gives a new insight into the physicochemical characteristics of the hot spots. The mean AUC of the classifier committee in 5-fold cross-validation was 0.89. The most balanced classifier, regarding the sensitivity S_n and specificity S_p , enabled the predictions with $S_n = 0.75, S_p = 0.87$, and AUC = 0.89.

The predictor of amylogenicity, called AmyloGram, is accessible as a web-server (ADRESS).

References

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