signalHsmm - a novel semi-Markov model of signal peptides

Michał Burdukiewicz[1], **Piotr Sobczyk**[2], **Paweł Błażej**[1], **and Paweł Mackiewicz**[1]

[1]**University of Wrocław, Department of Genomics, Poland**
[2]**Wrocław University of Technology, Department of Mathematics, Poland**

ABSTRACT

The proper localization of proteins in a cell is essential to perform their desired function. Information about the protein destination is included within the very protein in the form of short peptides called targeting signals. Ones of them are signal peptides, diverse N-terminal sequences, which are responsible for targeting of proteins to endomembrane system and their export outside the cell. Proteins equipped with signal peptides constitute a substantial fraction of the whole proteome and play crucial roles in metabolism, maintenance of tissue structure, immune response and regulation of other organismal functions. Moreover, the transport of proteins through the endomembrane system is important for their correct folding and posttranslational modifications. A common model of classical signal peptides assumes that they start with a positively charged n-region, followed by a hydrophobic h-region and a c-region ended with a cleavage site recognised by a signal peptidase.
However, our studies of many protein sequences representing the wide range of diversified taxonomic organisms indicate a variability of signal peptides. Therefore, the main goal of the study is to design a new probabilistic model for signal peptides, which will include knowledge about their organization, amino acid composition and variation.
The proposed model will be based on hidden semi-Markov models (HSMMs) and use intrinsic knowledge about signal peptides. The big advantage of the algorithm is a possibility to incorporate unique regions in the architecture of signal peptides. Therefore, HSMMs can be useful in recognition of atypical and artificial signal peptides. An aggregation of amino acids into physicochemical groups using the neural gas algorithm with k-fold cross-validation will reduce dimensionality of the problem and enable to learn the algorithm even on smaller data sets. The reduction will allow to use more complex probabilistic models and models specialized in the detection of signal peptides in different taxonomic groups of organisms, which improve the efficiency of the whole predictor. The n-gram and position-specific weight matrix methods will add more details to the probabilistic model.
Our preliminary model has showed the largest AUC=0.98 in comparison to other software and appeared very stable in the recovery of signal peptides after training even on very small data sets. Thanks to that, our model does not need to be permanently retrained with the continuous expansion of sequence databases. It should be emphasised that our model describes signal peptides from medically significant malaria parasites Plasmodium and their relatives (AUC = 0.92) more accurately than popular programs (0.84).

Keywords: Keyword1, Keyword2, Keyword3

INTRODUCTION

Your introduction goes here. Some examples of commonly used LaTeX commands and features are listed below, to help you get started.

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SOME LaTeX EXAMPLES

Use section and subsection commands to organize your document. LaTeX handles all the formatting and numbering automatically. Use ref and label commands for cross-references.

Figures and Tables

Use the table environment and the tabular command for basic tables — see Table 1, for example.

To include a figure in your document, use the figure environment and the includegraphics command as in the code for Figure 1.

Figure 1. An example image.

Item	Quantity
Widgets	42
Gadgets	13

Table 1. An example table.

Citations

LaTeX formats citations and references automatically using the bibliography records in your .bib file. Use the `\cite` command for an inline citation, like

Mathematics

LaTeX is great at typesetting mathematics. Let $X_1, X_2, \ldots, X_n$ be a sequence of independent and identically distributed random variables with $\mathrm{E}[X_i] = \mu$ and $\mathrm{Var}[X_i] = \sigma^2 < \infty$, and let

$$S_n = \frac{X_1 + X_2 + \cdots + X_n}{n} = \frac{1}{n}\sum_{i}^{n} X_i$$

denote their mean. Then as n approaches infinity, the random variables $\sqrt{n}(S_n - \mu)$ converge in distribution to a normal $\mathcal{N}(0, \sigma^2)$.

Lists

You can make lists with automatic numbering ...

1. Like this,

2. and like this.

. . . or bullet points . . .

- Like this,
- and like this.

. . . or with words and descriptions . . .

Word Definition

Concept Explanation

Idea Text

METHODS

Overview

The functionality of a signal peptide depends not on exact sequence of specific amino acids, but on the physicochemical properties of residues in a given region. Henceforth, the usage of raw amino acid sequences is superfluous and introduces unnecessary information. To utilize this property of signal peptide recognition, we cluster amino acids into several groups based on the physicochemical properties of residues.

The pre-processed sequences are further analyzed by the heuristic algorithm, which determines borders between three characteristic signal peptide regions, the enhanced version of algorithm presented in Nielsen and Krogh (1998). Using the current information from experimentalists, we refined the region recognition criteria.

Next, two models are trained to recognize proteins with and without a signal peptide. The first one is a hidden semi-Markov model, in which each of three signal peptide regions is represented by a different hidden state. The additional fourth hidden state represents mature protein. Each state is described by the frequencies of amino acid groups within that state. The distribution of hidden states durations, the number of amino acids related to each hidden state in signal peptide, is based on the empirical density of region lengths from the training set.

The second model is a simple probabilistic approach in which no association between amino acids was assumed, and probability of amino acids groups occurrence was determined by their frequencies in mature proteins.

Data selection

Eukaryotic protein sequences and their annotations were properly prepared according to the literature of the subject and downloaded from UniProt database release 2015_06. The positive set contained 2589 sequences with an experimentally confirmed signal peptide and its cleavage site. Sequences with more than one cleavage site were excluded from the final data set. The negative set comprised 152272 sequences without any signal peptide annotation. Protein sequences with ambiguous symbols: X, J, Z and B were removed from the final sets. Proteins with selenocysteine (U) were also excluded from data set, because there are no records of signal peptides containing this amino acid.

Clustering of amino acids

We cluster amino acids using several criteria relevant for the architecture of signal peptide: hydrophobicity, frequency in alpha-helices, polarity and size. High values of hydrophobicity are required in the h-region, the core of signal peptides. Alpha-helix, the secondary structure of this region, is probably induced by the positively charged n-region. Polarity as well as size are important especially in the cleavage site (Palzkill et al., 1994).

We selected 13 properties from AAIndex database (aaindexcitation) (see Table 2. Each property was attributed to a single criterion. Considering all combinations of single properties, we created 96 possible clusterings of amino acids using Euclidean distance and Ward's method.

Subsection

Here is an interesting equation that may be helpful in some situations:

$$\cos^3\theta = \frac{1}{4}\cos\theta + \frac{3}{4}\cos 3\theta \tag{1}$$

Paragraph Nothing to see here. Move on.

Criterion name	Property name
Size	Size
Size	Molecular weight
Size	Residue volume
Size	Bulkiness
Hydrophobicity	Normalized hydrophobicity scales for alpha-proteins
Hydrophobicity	Consensus normalized hydrophobicity scale
Hydrophobicity	Hydropathy index
Hydrophobicity	Surrounding hydrophobicity in alpha-helix
Polarity	Polarity
Polarity	Mean polarity
Frequency in alpha-helices	Signal sequence helical potential
Frequency in alpha-helices	Normalized frequency of N-terminal helix
Frequency in alpha-helices	Relative frequency in alpha-helix

Table 2. Properties used in clusterization.

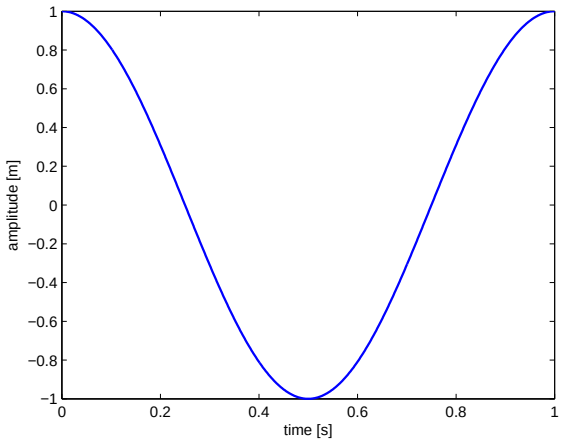

Figure 2. Can you guess which function this is?

Paragraph Really. See Figure 2 for more interesting results.

RESULTS AND DISCUSSION

You may want to separate results, discussion and conclusion, according to your needs.

Please submit the final pdf file via EasyChair to the GCB'15 program committee by June 30, 2015.

ACKNOWLEDGMENTS

Thank you for your support!

REFERENCES

Nielsen, H. and Krogh, A. (1998). Prediction of signal peptides and signal anchors by a hidden markov model. *Proceedings / ... International Conference on Intelligent Systems for Molecular Biology ; ISMB. International Conference on Intelligent Systems for Molecular Biology*, 6:122–130.

Palzkill, T., Le, Q. Q., Wong, A., and Botstein, D. (1994). Selection of functional signal peptide cleavage sites from a library of random sequences. *Journal of Bacteriology*, 176(3):563–568.