ConoDictor: a tool for prediction of conopeptide superfamilies

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ABSTRACT
ConoDictor is a tool that enables fast and accurate classification of conopeptides into superfamilies based on their amino acid sequence. ConoDictor combines predictions from two complementary approaches—profile hidden Markov models and generalized profiles. Results appear in a browser as tables that can be downloaded in various formats. This application is particularly valuable in view of the exponentially increasing number of conopeptides that are being identified. ConoDictor was written in Perl using the common gateway interface module with a php submission page. Sequence matching is performed with hmmsearch from HMMER 3 and ps_scan.pl from the pftools 2.3 package. ConoDictor is freely accessible at http://conco.ebc.ee.

INTRODUCTION
Conopeptides are the main bioactive component of cone snail venom. These marine animals produce complex venoms that contain hundreds of peptides and proteins. Recently, conopeptides have attracted a great deal of interest as a result of their selectivity for, and potent effects on, ion channels and receptors (1,2). Most are cysteine-knotted peptides that have been classified into superfamilies and families based on their structural or functional features (3,4). To date, >1500 non-redundant conopeptide sequences are stored in public databases and this number is increasing exponentially. Conopeptides are classified into ‘gene superfamilies’ based on their signal sequence. Currently, there are 16 major superfamilies, namely: A, D, I1, I2, I3, J, L, M, O1, O2, O3, P, S, T, V and Y. The precursors generally contain an N-terminal signal sequence, a central propeptide region and a C-terminal hypervariable mature toxin (4,5).

The Conoserver database (http://www.conoserver.org/) is a repository of nucleic acid and protein sequences, and of structural information on conopeptides (4).

The naming and classification of new conopeptide protein sequences has become an important issue because of the sharp increase in the number of new conopeptides being identified, and because studies to determine the peptide’s functional characteristics are based on this classification. The Conoserver prosequence analyzer (ConoPrec) is the most specific web tool available for elucidation of conopeptide class. It provides hints based on the signal peptide sequence of the submitted precursor (6). However, this tool does not work when the signal sequence is missing, which is often the case with conopeptides identified by proteomic and mass spectrometry studies of toxins identified as mature bioactive peptides in venom or dissected venom gland. As data generated by spreading venom high-throughput omics is notoriously incomplete, the classification of new sequences into conotoxin superfamilies should not be restricted to the signal peptide sequence. There is indisputable evidence for the relevance of consensus sequences of propeptides and cysteine frameworks in conopeptide sequences. Thus, the inclusion of these criteria should also be considered for conopeptide classification. We recently demonstrated the reliability of conopeptide family prediction and classification based on profile hidden Markov models (pHMM) of propeptides and mature peptides (7).

ConoDictor has been developed in the context of the CONCO project (www.conco.eu) and is a web-based tool that exploits pHMMs and position-specific scoring matrix (PSSM, also known as generalized profiles) to classify conopeptide into superfamilies based on their amino acid sequence. ConoDictor is a user-friendly tool that meets users’ demands for an easy-to-use environment for sequence classification and superfamily prediction. As a fully automated tool, ConoDictor provides classification results that must be checked by users.

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MATERIALS AND METHODS

Preparation of the model data set

Sequences used for generating the models were obtained from Conoserver. Only precursor sequences with gene superfamily annotation were considered. The training set consisted of 933 sequences. Each sequence was manually annotated with the gene superfamily classification after checking the classification provided by Conoserver. Each sequence was divided into three parts, which were stored separately: signal, propeptide and mature peptide. Separate files were also created for each of the 16 superfamilies. The sequences were then aligned using the MAFFT version 6.707b software. The alignments were manually refined when necessary using the JALVIEW 2.5 software, and the resulting 48 alignments were used to build the models.

Hidden Markov models

We previously described pHMM ability for conopeptide classification (7). We constructed pHMMs for each of the 48 alignments using the hmmbuild script from the HMMER 3.0 package (8,9). Matches between pHMMs and the sequence data set were searched using the hmmsearch script with an e-value significance level set to 0.1.

Generalized profiles (PSSM)

Generalized profiles were constructed using the pftool package, version 2.3. The most recent methodology based on annotated multiple sequence alignment (AMSA) was used. The generalized profiles were generated using apsimake in a semi-global mode after weighing of alignments. The resulting models were calibrated against randomized sequences and cut-off values tuned manually. These approaches have already been validated for classification of other proteins (10,11).

Testing of models on known conopeptides

The test set was constructed from publicly available conopeptide sequences extracted from the NCBI Protein database and UniProtKB (release 2010_11). The test set contained 1225 manually curated sequences. Sequences were manually annotated and assigned to the relevant superfamily according to UniProtKB annotations, cysteine frameworks and sequence similarity. Sequences not belonging to any superfamily were added to the test set as negative controls.

ConoDictor implementation

Input sequences are first classified using pHMMs and PSSMs separately. pHMM models of signal (X_sig), propeptide (X_pro) and mature peptide (X_mat) are used in parallel and corresponding predictions are combined. The same process is applied with PSSM models. Resulting pHMM and PSSM classifications are merged to produce a global combined classification.

RESULTS

Conopeptide models

For each of the 16 known conotoxin superfamilies, three separate models based on signal, pro- and mature peptides were built, providing a total of 48 hidden Markov models and 48 generalized profiles. The models were named according to the superfamily and the region of the precursor that they targeted. Each model demonstrated very
good discriminative abilities, with high sensitivity (≈95%) and selectivity (≈99%) ([7] and Koua et al., unpublished data). When tested using known conopeptide sequences, these models enabled extensive and reliable classification even between superfamilies containing mature peptides with high sequence similarities. The models provided good evidence of complementarity between signal, pro- and mature peptide sequences for superfamily determination, as well as complementarity between pHMMS and generalized profiles (Koua et al., submitted).

**ConoDictor input interface**

ConoDictor accepts amino acid sequences in FASTA format as input. The sequences can be pasted in the prepared field or uploaded as a file from the user’s computer (Figure 1). Sequences can be annotated with a predicted superfamily in the header between sharps (#), otherwise they are considered as ‘UNKNOWN’. By default, the models built in the framework of the CONCO project are used to analyse the input sequences. However, users can also upload their own PSSMs and/or pHMMs. An annotated testing set (attached to a ‘LOAD TEST DATA’ button) is also available from the input interface.

**Visualization interface**

The ConoDictor output interface offers user-friendly tab views of matching outputs and predictions (Figure 1). The main tab provides combined prediction, as well as a summary of pHMM- and PSSM-based prediction. Detailed result tabs for pHMM- and PSSM-based predictions provide the number of sequence matches for each model, the position for each sequence/model match, and the related e-value and score of individual model match. Tab headers and table column names explain the results displayed. The result page is automatically updated until analysis results are available. An Excel file (.xls) and raw text version (.txt) of all results can be downloaded. A session identifier is also provided, and the results can be accessed and visualized on the server for up to 3 weeks after submission or last viewing. A detailed help page provides clear explanations and screen shots of the most important tables of the analysis (http://conco.ebc.ee/ConoDictor_help.html).

**CONCLUSION**

ConoDictor is a web-based application, based on preliminary studies that established PSSM and pHMM complement each other for conopeptide identification and classification. Thanks to a user-friendly interface, ConoDictor provides an easy-to-use environment for classification of conopeptides into superfamilies based on their amino acid sequence. In view of the rapidly increasing number of new conopeptides being discovered...
by next-generation transcriptomic platforms, *ConoDictor* is a valuable bioinformatics tool for their classification and serves as a starting point for investigation of their functional characteristics.

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