PROFtmb: a web server for predicting bacterial transmembrane beta barrel proteins

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INTRODUCTION

Transmembrane beta-barrel (TMB) proteins form a beta-barrel as a single beta-sheet joined at its edges. The sheet is 'all-next-neighbor'(1), meaning all paired strands are adjacent in sequence. N- and C-termini of TMBs always reside in the periplasm. The architecture can be described as the repeating sequence. N- and C-termini of TMBs always reside in the membrane barrel, and a four-state per-residue labeling of upward- and downward-facing strands, periplasmic hairpins and extracellular loops. While most users submit individual proteins known to contain TMBs, some groups submit entire proteomes to screen for potential TMBs. Response time is about 4 min for a 500-residue protein. PROFtmb is a profile-based Hidden Markov Model (HMM) with an architecture mirroring the structure of TMBs. The per-residue accuracy on the 8-fold cross-validated testing set is 86% while whole-protein discrimination accuracy was 70 at 60% coverage. The PROFtmb web server includes all source code, training data and whole-proteome predictions from 78 Gram-negative bacterial genomes and is available freely and without registration at http://rostlab.org/services/proftmb.

PROCEDURE AND EXAMPLE OUTPUT

Users submit one or more FASTA-formatted protein sequences. For each sequence, PROFtmb builds a PSI-BLAST profile and runs the prediction, attempting to find the best fit of the protein to its TMB-based architecture, indicated as a Z-value. Results are always returned on a webpage, and take ~4 min per 500-residue protein. In the case of more than one sequence, an email of the results URL is sent.

If the query protein receives Z-value ≥ 4.0, PROFtmb provides a four-state (upward-strand, downward-strand, outer loop, periplasmic loop) per-residue prediction. Graphical output consists of color-coded four state posterior probability plots and amino acid sequence (Figure 1). Amino acid color indicates the final prediction, and usually corresponds to the state with maximum posterior probability, but with ‘corrections’ based on context shown with lighter-weight font [described in ‘Decoding’ section of the Supplementary Data of (2)]. While we did not quantify confidence levels for per-residue prediction, higher Z-values tend to have fewer corrected residues and greater contrast in state posterior probabilities.

In the example shown (Figure 1), OMPA from Escherichia coli [PDB: 1g90 (5) chain A] is predicted correctly at high confidence as an eight-stranded TMB. This result is expected, given PROFtmb was trained on very similar sequences. In most predictions on real TMBs, corrected residues are only

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found at the boundaries between strands and loops. Also, most strand and loop states have the best state close to probability 1.

In the second example shown (Figure 2), heme acquisition system protein A from *Serratia marcescens*, of the gammaproteobacteria class (Gram-negative) illustrates a false positive prediction. It receives a low but above-threshold Z-value of 4.8. In fact, the structure [PDB: 1B2V (6)] consists of a seven-stranded beta-sheet against four \( \alpha \)-helices. PROFtmb does correctly predict the locations of five of the strands. Notice that predicted strands four, five and six have poor contrast in posterior probability, indicating a poor fit to the PROFtmb model.

Finally, proteins shorter than 140 or longer than 1392 residues receive Z-value \(-10000\) (data not shown). The lower length of 140 is a conservative estimate of the smallest possible TMB, while the upper bound reflects the limit of our test set for Z-value calibration.

Occasionally, PROFtmb will assign Z-value less than four to a known TMB. Unfortunately, in such a case, the fact that it is a TMB can’t be used to help produce a reliable per-residue prediction since PROFtmb derives the prediction from sequence alone. This occurred in about 15% of the cases in our test set (see ‘Performance Evaluation’ in ‘Methods’ tab on website).

**DISCUSSION**

In our original paper (2) we used PSI-BLAST profiles run with options \(-h 1\) (\(E\)-value cutoff for inclusion in next pass) and \(-j 2\) (number of iterations), and did not explore the effect of different profiles on PROFtmb accuracy, either for whole protein or per-residue prediction. Since then, we have run 8-fold jackknife tests (leave one out, seven in) on the original SWISS-PROT sequence versions of eight PDB structures (SetTMBfull: 1a0s_P, 1af6_A, 1bt9_A, 1fep_A, 1prn, 1qd5_A, 1qj9_A, 1qjp_A). We built sets of PSI-BLAST profiles with 30 different combinations of settings \(-h \{1, 0.1, 0.01, 0.001, 0.0001, 0.00001\}\) and \(-j \{2, 3, 4, 5, 6\}\) and used each set in a separate jackknife test. The original Q2 accuracy, with settings \(-h 1 \ldots 2\) was 86.0%, while the best settings, \(-h 0.0001 \ldots 2\) achieved 87.3% Q2 accuracy. As a result, we changed the defaults to \(-h 0.0001 \ldots 2\). Additionally, we now allow the user to select these parameters. We have not estimated the effects of PSI-BLAST settings on whole-protein prediction yet. Currently, Z-value and resulting estimated accuracy and coverage are calibrated from our
original sequence-unique set called SetROC, containing a representative set of proteins from SWISS-PROT. As sequence databases are updated, we will periodically re-calibrate Z-values. A cluster plot and resulting accuracy versus coverage curve can be found in the ‘Methods’ section of the website.

DOWNLOADS

Predictions on 78 Gram-negative proteomes are available in the Download section, updated since original publication as follows. First, length-adjusted bits score was replaced by Z-value, which gives slightly improved discrimination on our test set (unpublished data). Second, per-residue predictions were re-run using updated PSI-BLAST profiles, with option /C0h 0.0001 rather than /C0h 1. Both changes are expected improvements, but haven’t been rigorously tested. Third, the model architecture now explicitly includes BEGIN and END states, representing the beginning and end of the amino acid sequence. This is required for the current version of the software.

The PROFtmb software is a general profile-HMM allowing specification of model architecture, encoding and decoding. The training data, consisting of eight TMB sequences with hand-annotated per-residue labeling based on their 3D structures, is available as well. Interested users may download and compile the C++ source code and use PROFtmb with the original training data or modify it. We make it available in the spirit of reproducibility, and encourage interested readers to contact the authors for more detailed advice.

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Conflict of interest statement. None declared.

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