iMembrane: homology-based membrane-insertion of proteins

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Abstract: iMembrane is a homology-based method, which predicts a membrane protein’s position within a lipid bilayer. It projects the results of coarse-grained molecular dynamics simulations onto any membrane protein structure or sequence provided by the user. iMembrane is simple to use and is currently the only computational method allowing the rapid prediction of a membrane protein’s lipid bilayer insertion. Bilayer insertion data are essential in the accurate structural modelling of membrane proteins or the design of drugs that target them.

Availability: http://imembrane.info. iMembrane is available under a non-commercial open-source licence, upon request.

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Supplementary information: Supplementary data are available at Bioinformatics online and at http://www.stats.ox.ac.uk/proteins/resources.

1 INTRODUCTION

Membrane proteins constitute ∼30% of all known proteins and are one of the largest classes of drug targets. They have roles in a multitude of biological processes such as cell recognition and neurotransmitter transport (Müller et al., 2008). Unfortunately, they are extremely hard to purify and crystalize, making experimentally determined structures rare. Current computational structure prediction methods are also not ideal, as they are designed to work on globular, soluble proteins.

However, even if a membrane protein’s structure is obtained, whether experimentally or computationally, we still do not hold the whole solution to the problem: the protein’s position within the lipid bilayer remains unknown. Natural ligands or drugs must be able to access the part of the protein to which they bind. Therefore, it is important to be able to distinguish the parts of the protein that are within the lipid bilayer from those that are solvent-accessible. This information is not currently available from experiments. Structures obtained by X-ray crystallography or nuclear magnetic resonance (NMR) spectroscopy do not reflect the protein’s native lipid bilayer environment.

There are several sequence-based methods to predict the position of transmembrane (TM) helices (e.g. TMHMM, Krogh et al., 2001) and β-barrels (e.g. HMM-B2TMR, Martelli et al., 2002). For reviews see Cuthbertson et al. (2005) and Bagos et al. (2005). The boundaries of putative TM helices or sheets tend to be predicted inaccurately and vary between different prediction methods. Half-helices, which span only a part of the membrane, are also hard to predict with existing tools. More importantly, all the above methods use a simple two-state membrane model (in membrane/not in membrane), occasionally with the addition of an uncertainty margin around the prediction. None of the available tools provides a detailed prediction of each residue’s position within the lipid bilayer, or its contacts with the different regions of the membrane lipids.

There are some structure-based methods, which predict a protein’s position within the membrane. These usually model the membrane as a hydrophobic slab, delimited by parallel planes (e.g. OPM, Lomize et al., 2006). The position of these planes is determined by using an energy function, which takes physical and/or statistical properties of amino acid residues as arguments.

In contrast to these largely simplified models, a recently developed method (Scott et al., 2008) uses coarse-grained molecular dynamics (MD) simulations in order to better account for the complexity of the lipid bilayer. Protein X-ray structures are simulated in the presence of membrane lipids, which self-assemble into a lipid bilayer. Simulation results include a summary listing the fraction of time each residue spent in contact with the different parts of the membrane lipids (polar head groups or hydrophobic lipid tails). A growing number of these simulation results are being made available online, in the Coarse-Grained database (CGDB, http://sbcb.bioch.ox.ac.uk/cgdb/). CGDB currently contains over 228 lipid bilayer self-assembly simulations for 138 PDB proteins covering 101 SCOP families, 90 superfamilies and 58 folds.

Performing MD simulations—even coarse-grained ones—requires large amounts of time and processing power. In this article, we present iMembrane, a simple method allowing the projection of the existing simulation results onto proteins of homologous structure or sequence. We show that these projected results do not vary greatly from those obtained in original coarse-grained simulations. Where performing an original simulation would take days on a compute server, our method takes mere seconds on a modern desktop computer. In addition, we are able to apply our method to proteins where only sequence information is available.

Here we use CGDB as our dataset. However, our method could theoretically be applied to any database of MD simulation results. Additional datasets will be included in future releases of iMembrane.

2 ALGORITHM

iMembrane accepts either a sequence, in FASTA format (Pearson, 1990), or a structure, in PDB format (Berman et al., 2000), as input.
In the case of a structure, its sequence is first extracted from the ATOM coordinates of the structure file. Typically, a BLAST (Altschul et al., 1990) sequence search is now carried out against the CGDB of membrane proteins. Matches are then realigned to the query using either MUSCLE (Edgar, 2004) sequence alignment or MAMMOTH (Ortiz et al., 2002) structure superposition. These alignments are then annotated using the CGDB protein’s simulation results. A flow diagram of the iMembrane algorithm, including alternative search methods, is available in the Supplementary Material.

A residue’s annotation is provided as a single letter per residue: N (not in contact with the membrane), H (in contact with the polar head groups of the membrane lipids) or T (in contact with the lipid hydrophobic tails). In the first instance, these letters simply represent an interpretation of the raw simulation results, as provided in the CGDB.

We also provide a simplified model, which abstracts the membrane as a three-layered slab, with an inner region around the membrane lipids’ hydrophobic tails, and two peripheral regions surrounding the membrane lipids’ polar head groups. The boundaries of these layers are calculated by fitting parallel planes onto the membrane contact data. This model allows us to then use each residue’s 3D coordinates to determine in which layer of the membrane it resides, or whether it is outside the membrane. iMembrane does this automatically for the CGDB proteins and then uses this information to annotate any homologous proteins aligned to them.

In the case where the input to our method is a structure, we can use the same procedure to assign every residue in the query protein to one of the membrane (or non-membrane) layers defined by the aligned CGDB protein. This step is performed in a Pymol environment (DeLano, 2002).

In the case of a sequence-only input, the query’s 3D information is missing. Therefore, we can only annotate those residues that are aligned to a CGDB protein’s residue. In the future, an additional structure prediction step will be implemented, such that we will be able to annotate every residue of a sequence-only input, as well as give back its proposed structure.

3 VISUALIZATION

We visualize the predicted membrane insertion of the input protein using (i) a colour-annotated sequence alignment and/or (ii) a coloured 3D structure as shown in Figure 1. The sequence-based visualization is always provided, whereas the coloured structure output is currently restricted to the case where the input itself was a structure.

4 ACCURACY

iMembrane’s accuracy was tested using a leave-one-out cross-validation on the CGDB data. The prediction results for each hit were compared to the original annotation generated directly from the corresponding MD simulation result in the CGDB. A Q3 score was calculated, representing the fraction of annotated residues assigned to the correct annotation (T, H or N; see Fig. 1). In addition, a Q2 score was calculated by merging the two types of membrane layers (T and H).

Independent of the input type (structure or sequence), a sequence identity of >35% tends to result in a Q3 accuracy >70% and a Q2 accuracy of ~90% and above in the membrane layer prediction. A slight upwards trend can be observed with increasing sequence identity. Below 35% sequence identity, homolog detection and sequence alignment quality is known to decline (Rost, 1999). As our method depends entirely on the alignment between the query and database proteins, its accuracy varies greatly below ~35% sequence identity, in the case where the input is a sequence. For structure input, this boundary is pushed down to 20% sequence identity. The use of improved alignment methods more suitable for distant homologs will benefit the accuracy of iMembrane in future releases.

A range of accuracy plots can be found in the Supplementary Material.

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