

Characterizing the transmembrane domains of ADAM10 and BACE1 and the impact of membrane composition

Conor B. Abraham,¹ Lin Xu,¹ George A. Pantelopulos,^{1,2} and John E. Straub^{1,*}

¹Department of Chemistry, Boston University, Boston, Massachusetts and ²Laboratory of Chemical Physics, National Institute of Diabetes and Digestive and Kidney Diseases, National Institutes of Health, Bethesda, Maryland

ABSTRACT The β -secretase, BACE1, and the α -secretase, ADAM10, are known to competitively cleave amyloid precursor protein (APP) in the amyloid cascades of Alzheimer's disease. Cleavage of APP by BACE1 produces a 99-residue C-terminal peptide (APP-C99) that is subsequently cleaved by γ -secretase to form amyloid- β (A β) protein, whereas cleavage of APP by ADAM10 is nonamyloidogenic. It has been speculated that ADAM10/APP and BACE1/APP interactions are regulated by colocalization within and outside of liquid-ordered membrane domains; however, the mechanism of this regulation and the character of the proteins' transmembrane domains are not well understood. In this work, we have developed and characterized minimal congener sequences for the transmembrane domains of ADAM10 and BACE1 using a multiscale modeling approach combining both temperature replica exchange and conventional molecular dynamics simulations based on the coarse-grained Martini2.2 and all-atom CHARMM36 force fields. Our results show that membrane composition impacts the character of the transmembrane domains of BACE1 and ADAM10, adding credence to the speculation that membrane domains are involved in the etiology of Alzheimer's disease.

SIGNIFICANCE Despite extensive research, the amyloid hypothesis, which posits that Alzheimer's disease begins with the production and aggregation of β -amyloid peptides, remains controversial. To settle this controversy, we need to better understand the molecular mechanisms governing β -amyloid production. It has been speculated that the localization of amyloid precursor protein, β -secretase, α -secretase, and γ -secretase into different membrane domains controls this process. Using molecular dynamics, we develop and characterize transmembrane domain models of the β -secretase, BACE1, and the α -secretase, ADAM10, in different membrane environments. Future studies can use these models to further develop our understanding of the molecular origins of Alzheimer's disease.

INTRODUCTION

The interplay between the β -secretase (β -site amyloid precursor protein cleaving enzyme 1; BACE1), the α -secretase (a disintegrin and metalloproteinase domain-containing protein 10; ADAM10), and amyloid precursor protein (APP) is a central focus in Alzheimer's disease (AD) research. According to the amyloid hypothesis of AD, misregulation of the competitive amyloidogenic and nonamyloidogenic APP processing pathways, mediated by BACE1, ADAM10, and γ -secretase, leads to an increase in amyloidogenic β -amyloid (A β) peptide production (1–3). The olig-

omerization and fibrilization of amyloidogenic A β has been identified as a key factor in the onset of AD (1–3).

The APP processing pathways begin with cleavage of APP by BACE1 at the β - or β' -sites of APP or by ADAM10 at its α -site (Fig. 1 A) (2). When APP is cleaved at its β -site, a 99-residue C-terminal transmembrane domain (TMD) fragment (APP-C99) is produced (2). Conversely, when cleaved at APP's β' -site or its α -site, a truncated C-terminal fragment is produced (2). Subsequent cleavage of APP-C99 and the truncated C-terminal APP fragments by γ -secretase produces amyloidogenic A β and nonamyloidogenic peptides, respectively (2). Despite extensive research, we do not fully understand what leads to the upregulation of amyloidogenic A β .

A possible explanation for the misregulation of these proteins could lie in the observation that high cholesterol is a

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*Correspondence: straub@bu.edu

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TMD sequences does not account for potential changes in membrane composition, and therefore it cannot provide a complete picture of the proteins' TMDs. Additionally, these predictions do not account for the role of the JMDs in the TMD's membrane insertion and character, which could be critical when developing a minimal model for TMD characterization.

In this work, we clearly define the TMD and JMD sequences of ADAM10 and BACE1 and develop congener models for this and future computational and experimental characterization. For each protein, we started with several lengths of peptides including residues in the predicted TMD region and within the flanking JMDs in an all- α -helical structure (Fig. 1 B and C). These peptides were then simulated using temperature replica exchange molecular dynamics (REMD) until a consistent helical region was identified and the conformational ensemble was adequately sampled. Once identified, the optimal length congener models were simulated in various membrane environments using conventional molecular dynamics with all-atom (CHARMM36) and coarse-grained (Martini2.2) force fields. Importantly, these models contain parts of the proteins' JMDs to preserve any JMD-membrane contacts that may be critical for replicating the properties of the TMDs. Due to the suggested role of membrane domains in mediating the interactions of ADAM10 and BACE1 with APP, we explored the effects of lipid environment on these congener models using simplified membrane models that approximately capture the lipid order of L_o and L_d domains. Dipalmitoylphosphatidylcholine (DPPC):cholesterol (CHOL) mixtures were the first binary mixture observed to produce the L_o lipid phase by Vist and Davis, and they have since been the principal binary lipid mixture used to represent the L_o phase (31). The L_o phase has frequently been used as a reductive representation of the lipid order expected in lipid rafts, which separates from the L_d phase, used to represent the lipid membrane bulk. As such, lipid phase separations that can be observed in models as simple as ternary mixtures of a saturated lipid, unsaturated lipid, and CHOL have frequently been used as simplified models of the formation and separation of lipid rafts from the lipid membrane bulk. In the molecular dynamics simulation literature, DPPC:dilinooleoylphosphatidylcholine (DIPC):CHOL has most-frequently been chosen as a ternary mixture for the study of lipid phase separation (6). As such, in selecting a simplified model of L_o and L_d lipid environments, we have chosen DPPC:CHOL and DIPC:CHOL binary lipid mixtures for characterization of the ADAM10 and BACE1 TMDs and JMDs.

Our results show that the proposed transmembrane congener models for ADAM10 and BACE1 are impacted by their membrane environment. By establishing minimal models for the TMDs of ADAM10 and BACE1, our study sets the stage for future computational and experimental work investigating the impact of membrane composition on ADAM10 and BACE1 structure and partitioning, as well as their interactions with the TMD of the APP substrate.

MATERIALS AND METHODS

Unless otherwise specified, simulation analysis was performed in Python3 with the Numpy, Scipy, Scikit-Learn, and/or MDAnalysis packages (32–36). All molecular visualizations were created using Visual Molecular Dynamics, Stride, and POV-Ray, and figures were created using Matplotlib (37–40). Probability density functions were computed using gaussian kernel density estimates with bandwidths optimized by 10-fold K-Fold cross validation (bandwidth values provided in Table S3). Further system and simulation details are provided in Tables S1 and S2.

All-atom proteins in implicit membranes

For each protein, transmembrane congener models with JMDs of varying length were simulated using REMD in implicit membranes with thicknesses of 30, 35, and 40 Å described by the generalized Born with simple switching (GBSW) potential in the CHARMM simulation program (41–43). This approach follows that taken in our previous work exploring the structure of APP-C99 (44). The congener models used for BACE1 included residues M440–K501, Q449–H490; and the congener models for ADAM10 included residues M643–P715, L654–L710, and E665–T703 (Fig. 1 B–C). Each model was constructed by placing the residues in a completely α -helical structure and centered in the GBSW membrane. The proteins were represented with the CHARMM36m force field (45). The GBSW membranes used a 6-Å smoothing length with 38 angular integration points and a maximum distance for radial integration of 20 Å. The nonpolar surface tension coefficient was set to 0.04 kcal/mol/Å². The membrane/solvent interface on each side of the membrane was set to 2.5 Å, leaving the hydrophobic cores of the 30-, 35-, and 40-Å membranes as 25, 30, and 35 Å, respectively. Replica exchange was attempted every 50 ps between 16 linearly spaced windows from 310K to 460K. Frames were saved every 10 ps for around 200 ns, and the frames from the last 100 ns from the 310K window were used for analysis.

Coarse-grained simulations

The Martini2.2 coarse-grained protein force field is dependent on a predefined secondary structure (46–48). To identify a suitable secondary structure, the final 100 ns of frames from the 310K window of the REMD simulations of the selected congener models were clustered with agglomerative clustering using Ward's minimum variance method (49). The optimal number of clusters was taken to be that with the highest Silhouette Score (50). This clustering scheme was applied using Julia with the Clustering.jl package and is further described in supporting methods I (51). For each protein and membrane width, the medoid of the largest cluster was mapped to the Martini2.2 model using the martinize.py program with a manually defined secondary structure (46–48). The secondary structures of the α -helical residues in the proteins were defined based on the averaged backbone hydrogen bond pattern of the frames in the largest cluster, and for the rest of the residues, they were defined using the DSSP result for the input structure (52).

For each protein, its Martini2.2 congener model was simulated in five single-component lipid membranes: 100 mol% DPPC, 100 mol% DIPC, 100 mol% dilauroylphosphatidylcholine (DLPC), 100 mol% dioleoylphosphatidylcholine (DOPC), and 100 mol% 1-palmitoyl-2-oleoylphosphatidylcholine (POPC). The models were also simulated in DIPC and DPPC membranes with four different concentrations of CHOL: 10 mol% CHOL, 20 mol% CHOL, 30 mol% CHOL, and 40 mol% CHOL. The congener model for each protein and membrane was selected based on the width of a pure lipid membrane composed of the lipid included in the membrane (Table S1). The systems were built using the insane.py program (53).

Five replicates were simulated for each system in GROMACS 2021.5 using the new-rf input parameters recommended by the Martini developers (54,55). Each replicate was first minimized by steepest descent and then

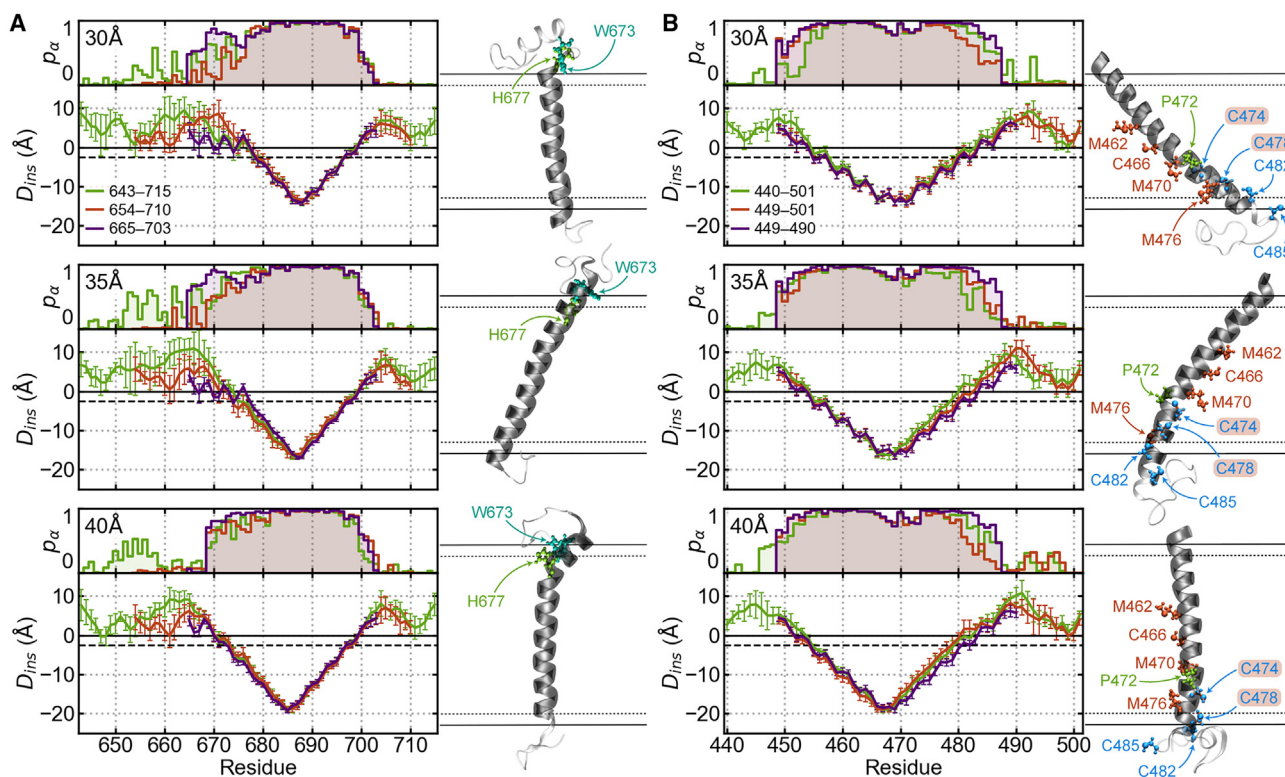


FIGURE 2 Structural comparison of congener models in implicit membranes. The residue α -helix propensity, p_{α} , and insertion depths, D_{ins} , are presented for (A) ADAM10 and (B) BACE1 TMD and JMD congener models in 30-Å (top), 35-Å (middle), and 40-Å (bottom) GBSW membranes. The insertion depth error bars indicate their standard deviations. The medoid structure of the largest cluster from the agglomerative clustering of each simulation is shown to the right of their D_{ins} and p_{α} results. For both proteins, the residues that induce a hinge/bend in the α -helix are green, and the extracellular JMD is oriented up. For BACE1, S-palmitoylatable residues are blue, and sulfur belt residues are orange.

equilibrated in the isobaric-isothermal (NPT) ensemble using the velocity rescaling thermostat and a semiisotropic Berendsen barostat with a compressibility of $4.5 \times 10^{-5} \text{ bar}^{-1}$ and an integration timestep of 20 fs (56). Production simulations were performed for 5 μs in the NPT ensemble using a semiisotropic Parrinello-Rahman barostat with a compressibility of $3.0 \times 10^{-4} \text{ bar}^{-1}$ (57). Coordinates were saved every 1 ns, and the last 4 μs of each simulation was used for analysis. All simulations were performed at 310K with an ion concentration of 0.15 M and with 10% of water represented with Martini's antifreeze water model.

All-atom proteins in explicit membranes

Final configurations from each 5- μs Martini2.2 replicate simulation of each protein in 7:3 DPPC:CHOL and 9:1 DIPC:CHOL membranes were backmapped to the CHARMM36 force field using the backwards.py program (58,59). Note that due to Martini2.2's one-to-four bead per atom representation, individual Martini lipids can represent multiple all-atom lipids. Here, the DIPC lipids were backmapped to di-C18:2 PC lipids and the DPPC lipids were backmapped to di-C16:0 PC lipids. Also note that di-C18:2 PC lipids are named DUPC in the CHARMM36 force field, but in this paper, we will continue to refer to them as DIPC for clarity. Backmapping was completed using a revised version of the intram-v5.sh program. The intram-v5.sh program is available with backwards.py on Martini's web site. The only revisions to intram were to use the Verlet cutoff scheme for neighbor searching and soft-core potentials to make it compatible with our single-precision version of GROMACS. Similar revisions are discussed and utilized in Martini's backmapping tutorials. An additional 2 nm of water was added along the z axis (perpendicular to the membrane) to the ADAM10 in the 9:1 DIPC:CHOL system using Packmol to prevent inter-

periodic contact between the JMDs (60). Each system was equilibrated for 5 ns in the NVT ensemble with the velocity rescaling thermostat and a 1-fs timestep and for 10 ns in the NPT ensemble with a 1-fs timestep, the velocity rescaling thermostat, and the semiisotropic Berendsen barostat with a compressibility of $4.5 \times 10^{-5} \text{ bar}^{-1}$ (56,61). Production simulations were performed using the Nose-Hoover thermostat and the semiisotropic Parrinello-Rahman barostat with a compressibility of $4.5 \times 10^{-5} \text{ bar}^{-1}$ for 1 μs (57,62,63). Coordinates were saved every 100 ps and the last 900 ns from each simulation was used for analysis. All simulations were performed using GROMACS 2021.2 at 310K with an ion concentration of 0.15 M (54,55).

RESULTS AND DISCUSSION

Development of BACE1 and ADAM10 TMD and JMD models

REMD simulations were performed to sample the conformational space of several length congener models of ADAM10 and BACE1 in 30-, 35-, and 40-Å GBSW implicit membranes. These simulations were used to identify minimal length congener sequences having consistent structural features in the three membranes to be used in explicit membrane simulations. Model validation principally focused on two metrics: residue α -helix propensity, p_{α} , and insertion depth, D_{ins} . Residue α -helix propensity was calculated as the probability that the i^{th} residue forms an amine-carbonyl

backbone hydrogen bond with residue $i + 4$ and/or residue $i - 4$. Hydrogen bonds were defined as having a donor-acceptor distance of ≤ 3.5 Å and a donor-hydrogen-acceptor angle $\leq 150^\circ$. Insertion depths were calculated as the nearest distance from the edge of the GBSW membrane.

For ADAM10, we evaluated model congener sequences including residues M643–P715, L654–L710, and E665–T703. The shortest model (E665–T703) features an extended α -helix and does not include reentrant residues within the N-terminal JMD (Fig. 2 A). We selected the L654–L710 model for further simulation to account for effects of interactions between the JMD and the membrane. In our REMD simulations of the ADAM10 congener models, residues H677 and W673 can form a sidechain-backbone hydrogen bond and induce a hinge in the TMD α -helix. From this point on, the L654–L710 model of ADAM10 will be referred to as ADAM10.

For BACE1, we evaluated three model congener sequences: M440–K501, Q449–K501, and Q449–H490. Helicity and insertion depths appear consistent across all three models in all membranes (Fig. 2 B). Therefore, the Q449–K501 model is the optimal model sequence, as including the last 11 residues of BACE1's C-terminal domain provided a more complete model congener sequence with minimal added computational cost. This BACE1 model contains four cysteine residues known to be S-palmitoylated (C474, C478, C482, and C485), five sulfur containing residues believed to contribute to BACE1 trimerization (M462, C466, M470, C474, and C478), and P472, which induces

a kink in the α -helix. From this point on, the Q449–K501 model of BACE1 will be referred to as BACE1.

Agglomerative clustering was performed on each REMD simulation, and the largest cluster was used to define the secondary structures of the selected models for Martini2.2 simulations (Figs. 2, 3, and 4). The medoid structures of the largest clusters (Fig. 2) were used as the input structure for mapping from all-atom to Martini representation.

Characterization of the ADAM10 TMD and JMD

The ADAM10 congener model was simulated in a variety of coarse-grained and all-atom lipid and lipid/CHOL bilayers. Insertion depths and α -helix propensities were calculated to determine the TMD residues. Insertion depths were calculated as the distance along the normal of the nearest leaflet between a residue's α -carbon (or BB bead for Martini) and the leaflet (supporting methods II). The position and normal of each leaflet were determined using the lipid phosphate phosphorus atoms within 2 nm of the any protein atom. Residue α -helix propensities were calculated in the same manner as with the REMD simulations. Here, we define the TMD as the continuous chain of residues passing through the bilayer with $D_{ins} \leq 0$ and the TMD α -helix as the continuous chain of residues with $p_\alpha \geq 0.5$.

To understand how the ADAM10 TMD and JMDs interact with and anchor to the membrane, the average number of hydrogen bonds, $\langle N_{HB} \rangle$, formed between each residue and each kind of membrane component were computed for each of the all-atom explicit membrane simulations.

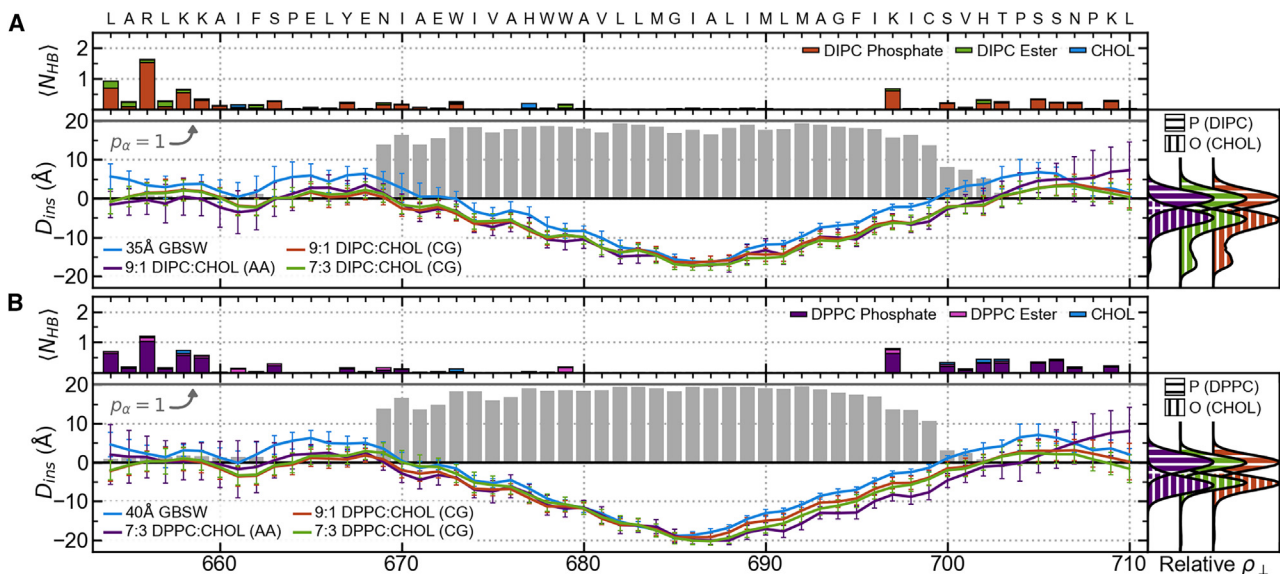


FIGURE 3 ADAM10 structure and interactions with membrane. Systems with (A) explicit DIPC membranes or a 35-Å implicit membrane and (B) explicit DPPC membranes or a 40-Å implicit membrane. For each set of membranes, residue insertion depths, D_{ins} , are shown with colored lines (error bars show standard deviations), and gray bars indicate the residues' α -helix propensities, p_α , for the all-atom explicit membrane system (bottom left). Relative lateral atomic densities along the membrane normal, ρ_\perp , for lipid phosphate phosphorus atoms and CHOL hydroxyl oxygen atoms (bottom right). Colored bars indicate the average number of hydrogen bonds formed between residues and membrane components ($\langle N_{HB} \rangle$) for the all-atom explicit membrane system (top).

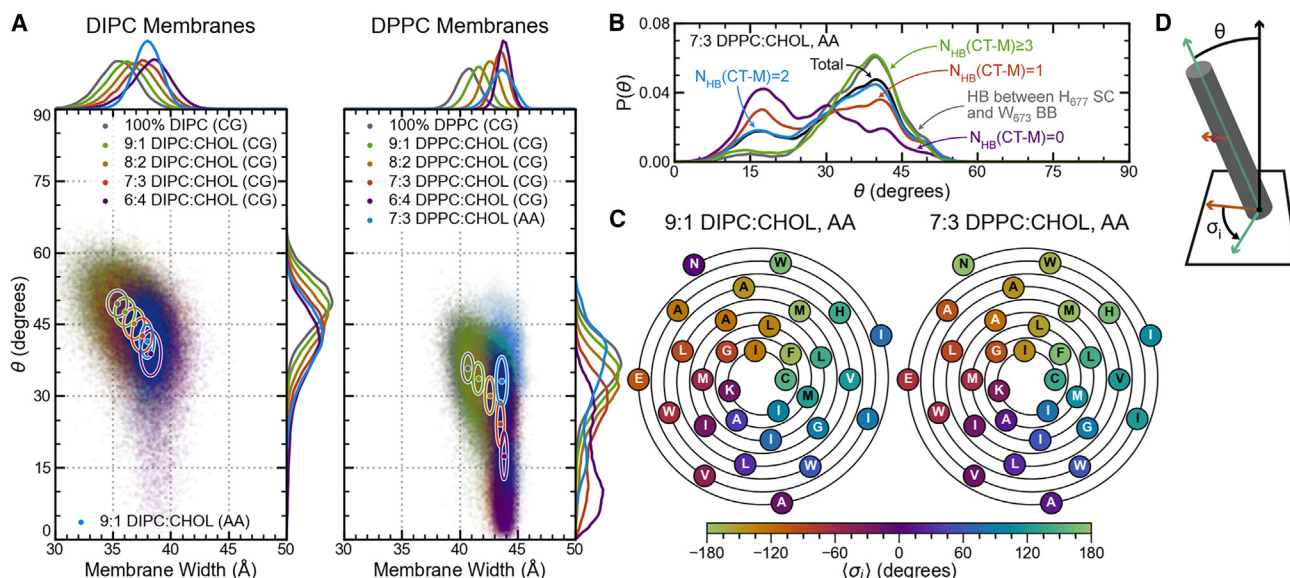


FIGURE 4 Impact of membrane composition on the orientation of ADAM10. (A) Effect of increase in membrane width on the tilt angle, θ , with confidence ellipses showing the standard deviations. Probability densities of θ and membrane width are shown along their corresponding axes. (B) Tilt angle probability density of all frames from all-atom simulation of ADAM10 in a 7:3 DPPC:CHOL bilayer compared with frames with a hydrogen bond between the sidechain (SC) of H677 and backbone (BB) of W673, and frames with different numbers of hydrogen bonds between the C-terminal JMD and the membrane, $N_{\text{HB}}(\text{CT-M})$. (C) Mean orientational angles, $\langle \sigma_i \rangle$, for each residue from all-atom simulations with 9:1 DIPC:CHOL and 7:3 DPPC:CHOL bilayers. (D) Reference graphic depicting θ and σ_i angles.

Potential hydrogen bond acceptors were identified as atoms whose charge $\leq -0.5 e$, potential hydrogen atoms involved in hydrogen bonds were identified as those with a charge $\geq 0.3 e$, and potential hydrogen bond donors were inferred from the potential hydrogens. Like with backbone hydrogen bonds, protein-membrane hydrogen bonds were defined as having a donor-acceptor distance of $\leq 3.5 \text{ \AA}$ and a donor-hydrogen-acceptor angle $\leq 150^\circ$.

For ADAM10 in a 9:1 DIPC:CHOL all-atom bilayer, the TMD extends from residues I670 to H702, and the α -helix includes residues N669–C699 (Fig. 3 A). In a 7:3 DPPC:CHOL bilayer, the ADAM10 TMD extends from residues I670 to P704, and residues N669–C699 make up the the α -helix. Similar results were found for the TMD in coarse-grained simulations with DIPC and DPPC membranes with various CHOL concentrations and in pure DOPC, DLPC, and POPC membranes (Fig. S5). These results show that the length of the ADAM10 TMD and TMD α -helix are not substantially impacted by membrane lipid and CHOL concentrations.

The tilt angle, θ , between ADAM10's α -helix axes and the membrane normal and the residues' mean orientational angles, $\langle \sigma_i \rangle$, were calculated for coarse-grained DIPC:CHOL and DPPC:CHOL systems with 0%, 10%, 30%, and 40% CHOL and for the all-atom systems with 9:1 DIPC:CHOL and 7:3 DPPC:CHOL membranes. The σ_i angles describe the orientation of a helical residue's α -carbon in the membrane about the axis of the α -helix, and the θ angle is measured as the angle between the α -helix axis and the membrane normal (Fig. 4 D). Further details about

the θ and σ_i calculations are provided in [supporting methods II](#).

For ADAM10 in DIPC membranes, the tilt angle decreases linearly with increasing membrane width. In contrast, for DPPC membranes with 30% and 40% CHOL (consistent with concentrations found in L_o membrane domains), ADAM10's tilt angle distribution becomes bimodal, indicating two distinct configurational states (Fig. 4 A). This bimodality also appears in all-atom 7:3 DPPC:CHOL membrane simulations of ADAM10 and can be accounted for by the number of hydrogen bonds between ADAM10's C-terminal JMD and membrane components and the presence of a hydrogen bond between the sidechain of H677 and the backbone of W673 (Fig. 4 B). Higher tilt angles tend to be associated with configurations with more than two hydrogen bonds between the C-terminal JMD and the membrane ($\bar{\theta} \approx 36^\circ$) and/or with a sidechain-backbone hydrogen bond between H677 and W673 ($\bar{\theta} \approx 37^\circ$). Lower tilt angles are most strongly associated with configurations lacking any hydrogen bonds between the C-terminal JMD and the membrane ($\bar{\theta} \approx 26^\circ$). We observe the impact of hydrogen bonds between the protein and the membrane on the tilt angle in the 7:3 DPPC:CHOL membrane but not in the 9:1 DIPC:CHOL membrane. This does not appear to be a product of hydrogen bonding with CHOL as, relative to lipid-protein hydrogen bonds, few of these interactions occur in either membrane (CHOL-protein hydrogen bonds account for only 5.7% and 9.6% of membrane-protein hydrogen bonds in the 9:1 DIPC:CHOL and 7:3 DPPC:CHOL membranes, respectively). Despite the observed change in tilt

angle between DIPC and DPPC membranes, the difference in $\langle\sigma_i\rangle$ from the 9:1 DIPC:CHOL all-atom to the 7:3 DPPC:CHOL all-atom membrane is unchanged ($10 \pm 36^\circ$; Fig. 4 C). On the N-terminus side, the TMD is anchored to the membrane primarily by residues L654, R656, and K658 (Fig. 3). On the C-terminus side, the TMD is anchored to the membrane by K697 and nonspecifically by various residues in the C-terminal JMD (Fig. 3).

To understand whether specific interactions with CHOL, other than just hydrogen bonding, could impact the character of ADAM10 in L_o domains, we analyzed residue-CHOL contacts with a residence time of greater than 25 ns in the all-atom 9:1 DUPC:CHOL and 7:3 DPPC:CHOL systems (Figs. S10A and S11). Within the C-terminal side leaflet, strong interactions occur between CHOL and F695. It is possible that in CHOL-rich L_o domains, these interactions lead to an unfavorable decrease in interactions between the C-terminal JMD and lipids and cause the previously described low-tilt state. Meanwhile in the N-terminal leaflet, interactions between H677, W678, and W679 and CHOL could impact the H677–W673 sidechain-backbone hydrogen bond, and interactions between F662 and CHOL could cause the slight decrease in hydrogen bonding between N-terminal JMD residues L654 and R656 observed in the 7:3

DPPC:CHOL system compared with the 9:1 DUPC:CHOL system.

Principal component analyses were performed for the all-atom 9:1 DIPC:CHOL and 7:3 DPPC:CHOL membrane simulations using the RMSD-fitted α -carbon cartesian positions of ADAM10. Principal component analyses have been applied for similar systems to identify critical global protein motions and can be easily applied using the Prody python package and other programs (64). Here, we instead used in-house code described in supporting methods III (65–67). Cross correlation matrices were constructed for each membrane using enough principal components to account for 90% of the cumulative variance. For the 9:1 DIPC:CHOL membrane and the 7:3 DPPC:CHOL membrane, the cumulative variances of the first 11 principal components are 90.8% and 91.2%, respectively. The cross correlation matrices of these principal components and the first two principal components for each system are shown in Fig. 5.

In both 9:1 DIPC:CHOL and 7:3 DPPC:CHOL membranes, the principal component analysis reveals strong correlations between helical residues and residues one helical turn later and a break in the helix between residues W673 and H677. This kink in the helix is more significant in the 7:3 DPPC:CHOL membrane than in the 9:1 DIPC:CHOL membrane. Further, in the 9:1 DIPC:CHOL membrane,

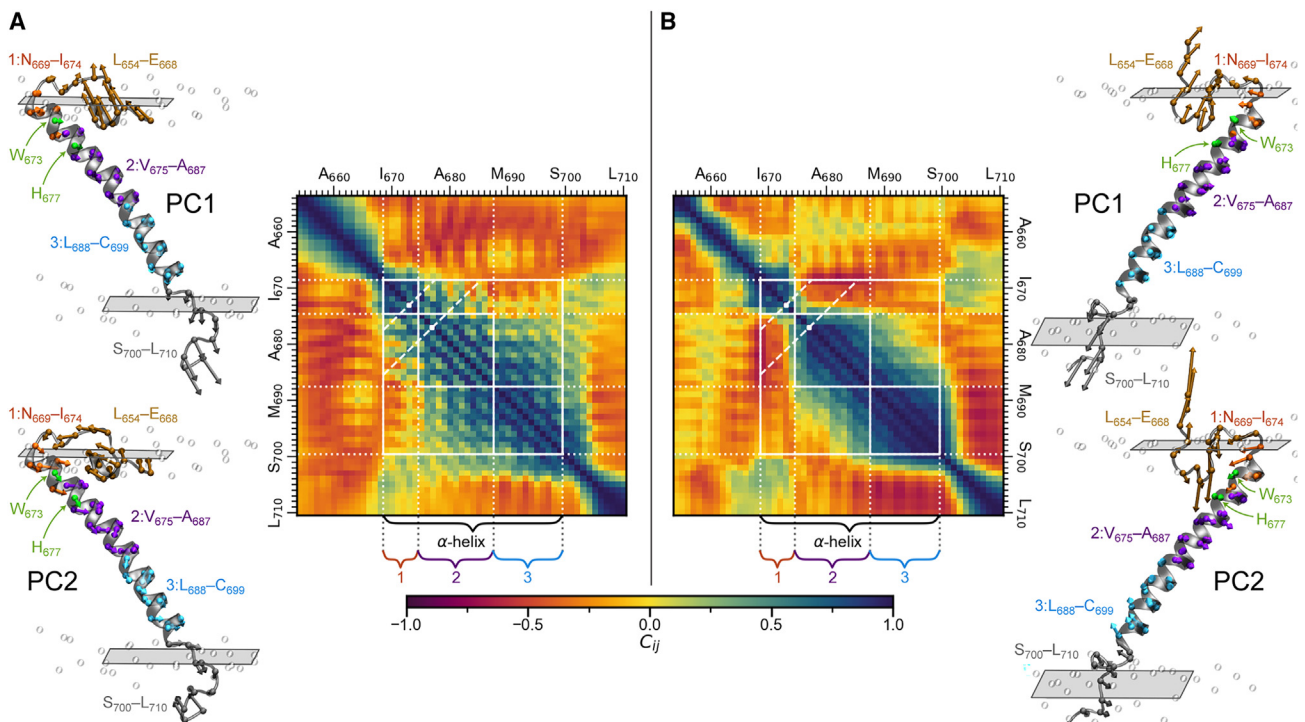


FIGURE 5 ADAM10 principal components and cross correlation matrices. The first (*top*) and second (*bottom*) principal components and principal component cross correlation matrices are shown for (A) 9:1 DIPC:CHOL and (B) 7:3 DPPC:CHOL all-atom membranes. The principal components are shown on the structure closest to the mean structure in PC1 and PC2 space. The lengths of the arrows are proportional to the magnitude of the principal components. N-terminus JMD residues are brown, C-terminus JMD residues are gray, W673 and H677 are green, and residue groups 1, 2, and 3 are orange, purple, and cyan, respectively. The cross correlation matrices were constructed using the first 11 principal components for each system to account for over 90% cumulative variance. Gridlines indicate the α -helix and residue groups 1, 2, and 3. Diagonal dashed lines mark the locations of W673 and H677.

residues in helical groups 2 (V675–A687) and 3 (L688–C699) are well correlated, whereas in the 7:3 DPPC:CHOL membrane, no significant correlation is observed. Finally, the principal component analysis further substantiates our previous TMD α -helix definition for ADAM10 of N669–C699 in both membranes.

Characterization of the BACE1 TMD and JMD

Following the protocol applied for ADAM10, the BACE1 congener model was simulated in a variety of coarse-grained and all-atom lipid and lipid/CHOL bilayers, and residue insertion depths and α -helix propensities were calculated. We define the TMD as the continuous chain of residues passing through the bilayer with $D_{ins} \leq 0$. The TMD α -helix was defined as the continuous chain of residues with $p_{\alpha} \geq 0.5$. To understand how BACE1's TMD and JMDs interact with the membrane, the average number of hydrogen bonds between each residue and each kind of membrane component were computed for each all-atom simulation. Hydrogen bonds were identified in the same manner as with ADAM10.

For BACE1 in a 9:1 DIPC:CHOL all-atom bilayer, the TMD extends from residues E452–C485 and the α -helix is made up of residues Q449–R484 (Fig. 6 A). In a 7:3 DPPC:CHOL all-atom bilayer, residues D451–Q479 make up the α -helix, and the TMD includes residues E452–R481 (Fig. 6 A). In this membrane, residues C482–L486 are also a part of the continuous chain of residues passing through the membrane with $D_{ins} \leq 0$. However, as they

are not a part of the α -helix and lie flat along the lipid headgroups, they are better classified as part of the C-terminal JMD. These all-atom results are in agreement with corresponding coarse-grained simulations with DIPC and DPPC membranes containing 0%, 10%, 20%, 30%, and 40% CHOL and in pure DOPC, DLPC, and POPC membranes (Fig. S6), suggesting that the length of BACE1's TMD and α -helix is dependent upon membrane composition due to structural changes in the membrane rather than specific interactions with individual membrane components.

The tilt angle, θ , and residue orientational angles, σ_i , were calculated for BACE1 in 9:1 DIPC:CHOL and 7:3 DPPC:CHOL all-atom bilayers, and θ was calculated for coarse-grained DIPC and DPPC bilayers with a variety of CHOL concentrations. For BACE1, the tilt angle decreases linearly with increasing membrane width for both DIPC and DPPC membranes (Fig. 7 A). BACE1's shorter TMD α -helix in DPPC membranes leads to a more pronounced decrease in tilt angle compared with DIPC membranes than what was observed for ADAM10 (Fig. 4 A). In both membranes, BACE1 is anchored to the membrane by hydrogen bonds between the membrane and R481, R484, and R487, belonging to an RCLRCLR motif. However, compared with in the 9:1 DIPC:CHOL membrane, hydrogen binding propensity with the membrane is shifted from R481 to R484 in the 7:3 DPPC:CHOL membrane (Fig. 6). This change in anchoring residue provides a plausible explanation for the shorter TMD α -helix observed in the wider 7:3 DPPC:CHOL membrane and results in a change in the residue orientational angles (Fig. 7 B). The

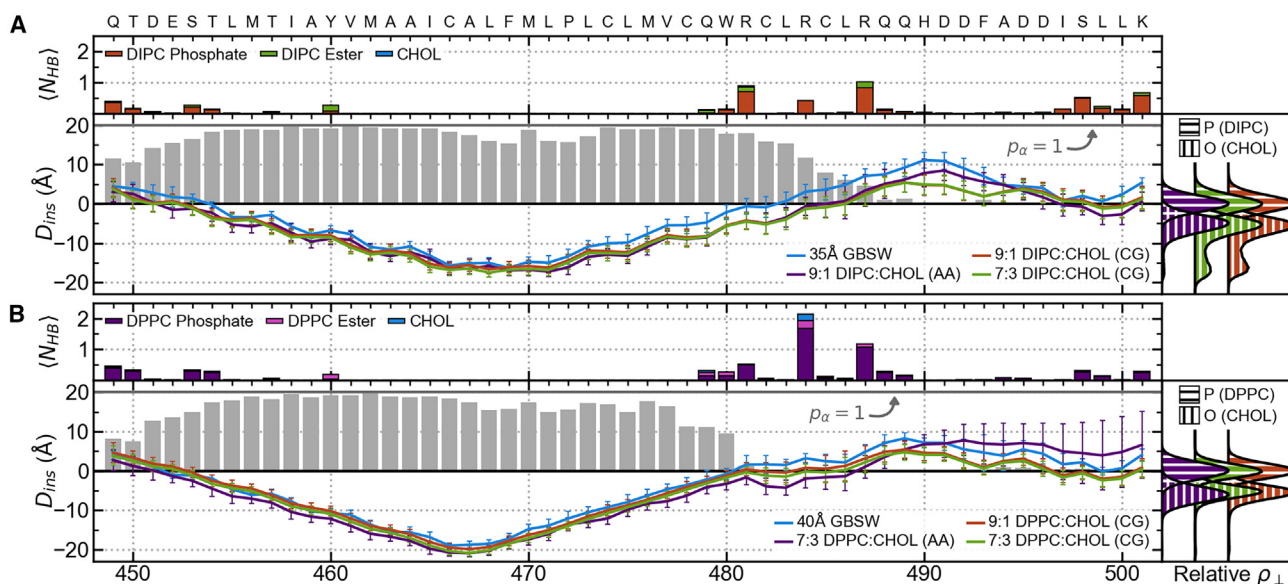


FIGURE 6 BACE1 structure and interactions with membrane. Systems with (A) explicit DIPC membranes or a 35-Å implicit membrane and (B) explicit DPPC membranes or a 40-Å implicit membrane. For each set of membranes, residue insertion depths, D_{ins} , are shown with colored lines (error bars show standard deviations), and gray bars indicate the residues' α -helix propensities, p_{α} , for the all-atom explicit membrane system (bottom left). Relative lateral atomic densities along the membrane normal, ρ_{\perp} , are shown for lipid phosphate phosphorus atoms and CHOL hydroxyl oxygen atoms (bottom right). Colored bars indicate the average number of hydrogen bonds formed between residues and membrane components (N_{HB}) for the all-atom explicit membrane system (top).

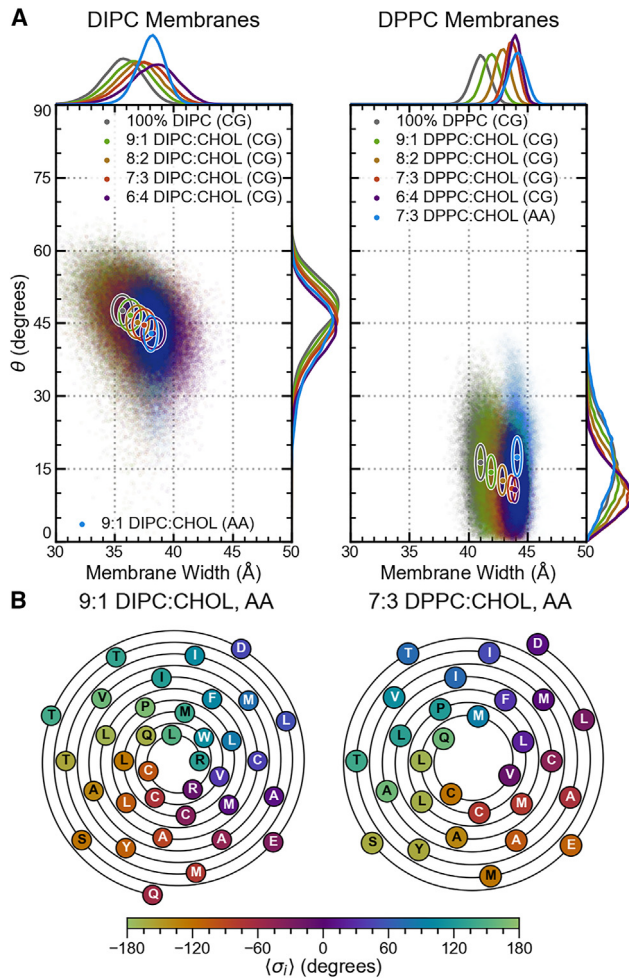


FIGURE 7 Impact of membrane composition on the orientation of BACE1. (A) Effect of increase in membrane width on the tilt angle, θ , with confidence ellipses showing the standard deviations. Probability densities of θ and membrane width are shown along their corresponding axes. (B) Mean orientational angles, $\langle\sigma_i\rangle$, for each residue from all-atom simulations with 9:1 DIPC:CHOL and 7:3 DPPC:CHOL bilayers.

average change in $\langle\sigma_i\rangle$ from the 9:1 DIPC:CHOL membrane to the 7:3 DPPC:CHOL membrane is $-58 \pm 13^\circ$, corresponding to the helical twist between the anchoring residues R481 and R484.

To further understand how CHOL impacts the character of BACE1, we examined specific residue-CHOL contacts lasting longer than 25 ns in the all-atom 9:1 DIPC:CHOL and 7:3 DPPC:CHOL systems (Figs. S10B and S12). Within the N-terminal leaflet, these contacts appear to be facilitated by Y460, whose α -carbon is positioned on the side of the helix pointing slightly into the membrane in the 9:1 DIPC:CHOL system ($\langle\sigma_i\rangle \approx -109^\circ$) and more significantly into the membrane in the 7:3 DPPC:CHOL system ($\langle\sigma_i\rangle \approx -161^\circ$). On the C-terminal side of the helix, protein-CHOL contacts appear to be facilitated by F469 and W480. In the 9:1 DIPC:CHOL membrane, F469 ($\langle\sigma_i\rangle \approx 100^\circ$) and W480 ($\langle\sigma_i\rangle \approx 107^\circ$) are both on the side of the helix pointing

slightly toward the outside of the membrane. In the 7:3 DPPC:CHOL system, F469 points in toward the center of the membrane ($\langle\sigma_i\rangle \approx 35^\circ$), and W480 becomes a part of the C-terminal JMD. It is possible that the interactions between these residues and CHOL drive the reorientation of BACE1 in L_o domains.

Principal component analyses were performed for both all-atom systems using the RMSD-fitted α -carbon cartesian positions of BACE1. Cross correlation matrices were constructed using the first 11 principal components from the 9:1 DIPC:CHOL system (90.5% cumulative variance) and the first 10 principal components from the 7:3 DPPC:CHOL system (90.5% cumulative variance). The cross correlation matrices and the first two principal components for each system are shown in Fig. 8. The cross correlation matrices reaffirm our previous observation of a shortened helical structure in the 7:3 DPPC:CHOL bilayer compared with the 9:1 DIPC:CHOL bilayer. In the 7:3 DPPC:CHOL bilayer, group 1 (Q449 and T450) and group 4 (W480–C485) lack the helical character seen in the 9:1 DIPC:CHOL bilayer. In the 7:3 DPPC:CHOL bilayer, the essential motions of the residues from L483 to Q488 are better correlated with one another than they are in the 9:1 DIPC:CHOL membrane. This is a consequence of having a membrane anchor within the C-terminal side of the TMD α -helix and another within the C-terminal JMD.

CONCLUSIONS

Our findings show that ideal transmembrane congener models of ADAM10 and BACE1 include TMD and JMD residues L654–L710 and Q449–K501, respectively. These models were used in our characterization of each protein's TMD and JMDs and are aptly suited for future computational and experimental characterization.

In both a 9:1 DIPC:CHOL bilayer and a 7:3 DPPC:CHOL bilayer, ADAM10's TMD α -helix includes residues N669–C699. ADAM10's TMD is found to be consistently bound to the membrane by hydrogen bonds between lipids and residue K697 on the C-terminal side of its TMD and residues L654, R656, and K658 on the N-terminal side of its TMD. The ADAM10 congener model also interacts with the membrane through protein-CHOL contacts, involving aromatic residues H677, W678, W679, F662, and F695, which are observed to increase in the CHOL-rich 7:3 DPPC:CHOL membrane. With a consistent α -helix definition and anchoring residues across different membranes, the TMD's orientation in the membrane about its α -helix axis did not differ between the 9:1 DIPC:CHOL membrane and the 7:3 DPPC:CHOL membrane. In membranes with compositions consistent to those found in L_d membrane domains, the tilt of the α -helix decreases linearly with increasing membrane width. However, a second configurational state with a lower tilt angle arises in L_o membranes. This configurational state tends to lack a backbone-sidechain hydrogen bond between

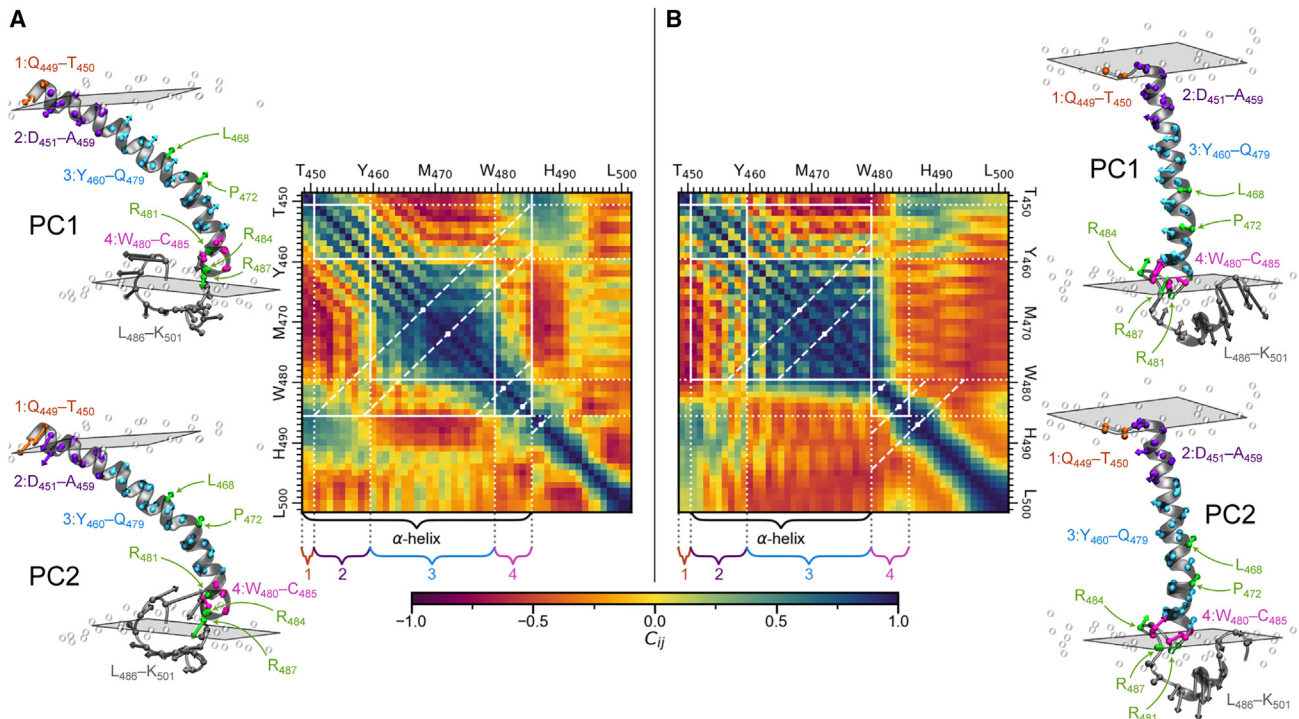


FIGURE 8 BACE1 principal components and cross correlation matrices. The first (*top*) and second (*bottom*) principal components and principal component cross correlation matrices are shown for (A) 9:1 DIPC:CHOL and (B) 7:3 DPPC:CHOL all-atom membranes. The principal components are shown on the structures closest to the mean structure in PC1 and PC2 space. The lengths of the arrows are proportional to the magnitude of the principal components. C-terminus JMD residues are gray, L468, P472, R481, R484, and R487 are green, and residue groups 1, 2, 3, and 4 are orange, purple, cyan, and pink, respectively. To account for over 90% cumulative variance, the cross correlation matrices for 9:1 DIPC:CHOL and 7:3 DPPC:CHOL were constructed using the first 11 and 10 principal components, respectively. Gridlines indicate the α -helix, and residue groups 1, 2, 3, and 4. Diagonal dashed lines mark the locations of L468, P472, R481, R484, and R487.

W673 and H677 and typically has few hydrogen bonds binding the C-terminal JMD to the membrane. This behavior in L_o membranes could explain the speculated localization and enzymatic activity of ADAM10 in L_d domains, but this requires further investigation.

In both 9:1 DIPC:CHOL and 7:3 DPPC:CHOL bilayers, BACE1 is primarily bound to the membrane by an RCLRCLR motif on the C-terminal side of the TMD and in the C-terminal JMD. However, in the 9:1 DIPC:CHOL membrane, R481 and R487 are the primary anchors, whereas in the 7:3 DPPC:CHOL bilayer, R484 is the primary anchor. Despite having a wider membrane width, this results in a shorter TMD α -helix in the 7:3 DPPC:CHOL bilayer (D451–Q479) than in the 9:1 DIPC:CHOL bilayer (E449–C484). This shorter α -helix results in a more substantial decrease in tilt angle in DPPC membranes than in DIPC membranes compared with what was seen with ADAM10. The BACE1 congener model also interacts with the membrane through protein-CHOL contacts, involving aromatic residues Y460, W480, and F469, which are observed to increase in the CHOL-rich 7:3 DPPC:CHOL membrane. The change in anchoring residues causes a rotation of the TMD α -helix about its axis in the membrane. Together, these factors could contribute to BACE1's parti-

tioning, homodimerization, and enzymatic activity. It is important to note that our simulations were not of palmitoylated BACE1, which may be necessary for proper partitioning and activity.

It is notable that both ADAM10 and BACE1 have arginine residues within their TMD and JMD sequences, especially BACE1 with its RCLRCLR anchoring motif. Arginine-rich sequences are also observed in helical cell-penetrating peptides that can readily insert into membranes and be internalized by the cell (68). It is possible that the prevalence of arginine in and near the TMDs of ADAM10 and BACE1 facilitates their membrane insertion and trafficking. Despite a large body of research focusing on ADAM10 and BACE1, we still do not know whether the proteins cleave APP in L_o domains or in the L_d domain. The principal impact of lipid composition on the structure and orientation of the BACE1 and ADAM10 TMDs results from variation in bilayer thickness, membrane ordering, and specific interactions with CHOL. In the case of BACE1, this impacts the length of the TMD α -helix and the tilt and orientation of the helix in the membrane. Thanks to its RCLRCLR motif, BACE1 is able to more readily adapt to the wider L_o membranes than ADAM10. For ADAM10, the wider membranes only cause a change in its tilt angle, and in L_o membranes, only part of

its ensemble adopts a substantially lower tilt conformation. This suggests ADAM10 may be less suited for localization in L_o domains than BACE1. These effects of membrane composition could impact dimerization of the proteins with APP, preventing or augmenting their APP-cleaving activity. Although this work does not address the question of where ADAM10 and BACE1 cleave APP directly, it provides insight into the influence of membrane composition on the structures and fluctuations of the TMDs of ADAM10 and BACE1. With this foundation of knowledge, the TMD models developed here can be used in future studies exploring their partitioning between L_o and L_d membrane domains and detailed interactions with APP believed to be essential elements of the biogenesis of A β and the amyloid cascades of AD. Frames from the all-atom explicit membrane systems for each protein congener model are provided in the [supporting material \(Data S1–S4\)](#) for the reader's convenience.

SUPPORTING MATERIAL

Supporting material can be found online at <https://doi.org/10.1016/j.bpj.2023.08.025>.

AUTHOR CONTRIBUTIONS

This research was designed by C.B.A., L.X., G.A.P., and J.E.S. Implicit membrane with all-atom protein REMD simulations were performed by L.X. and G.A.P. and analyzed by L.X., G.A.P., and C.B.A. Coarse-grained and all-atom conventional MD simulations were performed and analyzed by C.B.A. The article was written by C.B.A., G.A.P., and J.E.S.

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DECLARATION OF INTERESTS

The authors declare no competing interests.

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