

Supramolecular organization and dynamics of mannosylated phosphatidylinositol lipids in the mycobacterial plasma membrane

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1 Supramolecular Organisation and Dynamics of 2 Mannosylated Phosphatidylinositol Lipids in the Mycobacterial Plasma 3 Membrane 4 Chelsea M. Brown¹, Robin A. Corey², Axelle Grélard³, Ya Gao^{4,5}, Yeol Kyo Choi⁵, Emanuel Luna⁵, 5 Martine Gilleron⁶, Nicolas Destainville⁷, Jérôme Nigou⁶, Antoine Loquet³, Elizabeth Fullam¹, Wonpil Im^{5,*}, Phillip J. Stansfeld^{1,8,*,+}, Matthieu Chavent^{6,*,+} 6 7 8 ¹ School of Life Sciences, 9 University of Warwick, 10 Coventry, 11 CV4 7AL, 12 UK 13 14 ² Department of Biochemistry, 15 University of Oxford, 16 Oxford, 17 UK 18 19 ³ Université de Bordeaux, 20 CBMN (UMR5248), 21 CNRS, IPB, 22 Institut Européen de Chimie et Biologie, 23 F-33600 Pessac, 24 France 25 26 ⁴ School of Mathematics, Physics and Statistics, 27 Shanghai University of Engineering Science, 28 Shanghai 201620, 29 China 30 ⁵ Department of Biological Sciences, 31 32 Department of Chemistry, 33 Department of Bioengineering, 34 Lehigh University, 35 Pennsylvania 18015, 36 USA 37 ⁶ Institut de Pharmacologie et Biologie Structurale (IPBS), 38 39 Université de Toulouse, CNRS, 40 Université Toulouse III - Paul Sabatier,

42 France 43 44 ⁷ Laboratoire de Physique Théorique, 45 Université de Toulouse, CNRS, 46 UPS, Toulouse, 47 France 48 49 ⁸ Department of Chemistry, 50 University of Warwick, 51 Coventry, 52 CV4 7AL, 53 UK 54 55 + equal contribution 56 57 *to whom correspondence should be addressed: 58 e-mail: wonpil@lehigh.edu 59 phone: +1 (0) 610-758-4524 60 61 e-mail: phillip.stansfeld@warwick.ac.uk 62 +44 (0) 2476523864 phone: 63 64 e-mail: matthieu.chavent@ipbs.fr 65 phone: +33 (0) 5-61-17-59-00 66

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Abstract

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Mycobacterium tuberculosis (Mtb) is the causative agent of tuberculosis (TB), a disease that claims ~1.6 million lives annually. The current treatment regime is long and expensive, and missed doses contribute to drug resistance. Therefore, development of new anti-TB drugs remains one of the highest public health priorities. Mtb has evolved a complex cell envelope that represents a formidable barrier to antibiotics. The Mtb cell envelop consists of four distinct layers enriched for Mtb specific lipids and glycans. Although the outer membrane, comprised of mycolic acid esters, has been extensively studied, less is known about the plasma membrane, which also plays a critical role in impacting antibiotic efficacy. The Mtb plasma membrane has a unique lipid composition, with mannosylated phosphatidylinositol lipids (phosphatidyl-myoinositol mannosides, PIMs) comprising more than 50% of the lipids. However, the role of PIMs in the structure and function of the membrane remains elusive. Here, we used multiscale molecular dynamics (MD) simulations to understand the structure-function relationship of the PIM lipid family and decipher how they self-organize to shape the biophysical properties of mycobacterial plasma membranes. We assess both symmetric and asymmetric assemblies of the Mtb plasma membrane, and compare this with residue distributions of Mtb integral membrane protein structures. To further validate the model, we tested known anti-TB drugs and demonstrated that our models agree with experimental results. Thus, our work sheds new light on the organization of the mycobacterial plasma membrane. This paves the way for future studies on antibiotic development and understanding Mtb membrane protein function.

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Significance statement

Understanding the biophysical properties of mycobacterial membranes is of crucial interest, especially in the context of drug discovery, but there is no realistic model of mycobacterial membranes. In this work, we have developed models of inner membrane of mycobacteria mainly constituted of mannosylated phosphatidylinositol lipids. Our results provide significant and unexpected insights into the molecular details of mycobacterial membranes. Especially, they highlight how recently determined proteins can accommodate into this specific membrane and how its lipid composition can affect antibiotics diffusion.

Introduction

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Tuberculosis (TB) is caused by Mycobacterium tuberculosis (Mtb). In 2021 alone, there were an estimated 10.6 million new *Mtb* infections, leading to 1.6 million deaths (1). Thus, *Mtb* is one of the world's leading infectious killers, despite the availability of both a treatment regime and vaccine. The current course of antibiotics for drug susceptible TB cases can last as long as 6 months and consists of four drugs given in combination (2). This is not only expensive and demanding for the patient, but also encourages non-compliance that contributes towards the rise in multi-drug resistant and extremely drug-resistant TB (3). It is obvious that new treatments and a better vaccine are needed to meet the World Health Organization's 'End TB Strategy'. Their plan aims to reduce TB-related deaths by 90% by 2030, and thereby curtail the enormous public health cost caused by TB (4). The COVID-19 pandemic has undone some of the progress that had been made in the treatment of TB as fewer people were able to be diagnosed or access medication (5), which might result in a worldwide surge of untreatable cases. Thus, there is a pressing need for innovative research into the mechanisms of Mtb virulence and its ability to survive within the host for extended periods to help develop alternative intervention strategies.

One issue in anti-TB drug discovery is the complexity of the mycobacterial cell envelope (6, 7). This cell envelope consists of an array of lipids contributing to both hydrophobic and polar regions of various thicknesses and densities, making it extremely challenging to predict how molecules will cross this barrier and enter the cell. The Mtb cell envelope has four distinct layers: the outer layer (or the mycomembrane) comprised of mycolic acid esters and other complex lipids (such as dimycoserosates), an phthiocerol arabinogalactan-peptidoglycan layer, periplasmic space and the inner membrane, or the plasma membrane (8). The mycobacterial plasma membrane plays a key role in controlling nutrient/antibiotic uptake and contains important membrane proteins that are targets for antitubercular drugs (9), such as SQ109 that inhibits the transporter MmpL3 (10). The membrane composition is also known to change during different growth stages of the bacteria (7) and may modulate drug and membrane protein diffusion. Additionally, recent studies showed that Mtb plasma membrane organization, such as the formation of functional membrane microdomains (11, 12), can affect the survival ability of mycobacteria. Therefore, understanding the molecular organization and dynamics of the *Mtb* plasma membrane is essential for developing effective drug candidates.

The mycobacterial plasma membrane is composed of a variety of lipids and glycolipids, the most abundant being (Figure **1A**): cardiolipin (CL), phosphatidylethanolamine (PE), phosphatidyl-myoinositol (PI), trehalose monomycolate (TMM) and phosphatidyl-myoinositol mannosides (PIMs) (13), with the PIMs accounting for over half the dry weight of the plasma membrane lipids (13). PIM₂ lipids are comprised of a modified PI core decorated with two mannose residues and one acyl chain (Figure 1B). Furthermore, additional modifications of an acyl group and up to 4 mannose sugars can be added to the core headgroup, further diversifying the PIM structure (14, 15) (Figure 1B and SI Figure 1). It is still unclear if the plasma membrane is symmetric or asymmetric (16). In one widely cited and accepted study, it is proposed to be asymmetric (13), with Ac₂PIM₂ being the dominant species in the cytoplasmic leaflet accompanied by AcPIM₂, while the periplasmic leaflet is more varied, containing AcPIM₂, AcPIM₆, Ac₂PIM₆, CL, PI, PE and TMM (6, 13, 17, 18). PIM₆ lipids can be further modified to LM and LAM that make up the bulk of the periplasmic space (19). Additionally, lipids have been shown to modulate membrane protein function (20, 21). However, due to the complexity of the mycobacterial cell envelope, the dynamics and properties of the plasma membrane is extremely difficult to probe experimentally.

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Over the last decade, molecular dynamics (MD) simulations have emerged as a powerful strategy for studying structural and functional aspects of biomembranes (22, 23). However, at the time of writing, there are no complete models of the mycobacterial plasma membrane that capture a possible asymmetry and the diverse range of lipids (24). This limits our ability to understand *Mtb* plasma membrane biology, model drug-lipid interactions - important for anti-TB drug discovery efforts - and simulate mycobacterial membrane proteins in a native environment. Therefore, it is essential to develop models of the major phospholipids of the mycobacterial plasma membrane for application in MD simulations.

Herein, multiscale simulations were used to analyze the structure-function relationship of the four main PIM lipids (AcPIM₂, Ac₂PIM₂, AcPIM₆ and Ac₂PIM₆) found in the mycobacterial membrane. An asymmetric bilayer containing these lipids was assembled and simulated, showing the stability of this composition. To highlight the robustness of our approach, additional simulations representing (in total) two different asymmetric growth stages as well as symmetric membrane configurations were performed. Systematic analysis of the distribution of *Mtb* protein residues in contact

with lipids reveals further indications that this membrane could possibly be asymmetric. We have also modelled how antibiotics, used to treat TB infections, diffuse through the membrane and interact with membrane proteins and demonstrated that our models agree with experimental results (25). Overall, we have developed robust possible representations of the mycobacterial plasma membrane that will enable studies of membrane dynamics, lipid interactions with integral membrane proteins, and diffusion of antibiotics across this barrier.

Methods

Building the coarse-grained lipid parameters

The CG models of the lipids were parametrized for the newly released MARTINI 3 force field (26) and generated using the protocol described for small molecules (27) on the MARTINI website (http://cgmartini.nl/index.php/martini-3-tutorials/parameterizing-a-new-small-molecule). The bead types and mapping to the PIM molecules were performed manually, comparing with the recently published CG model of PI (28). Atoms were grouped according to functional groups, in sets of 3-5 non-hydrogen atoms. The CG mapping of Ac_xPIM_x is shown in **Figures 1B,C**. The parameter files that describe the bonds, constraints, and angles were assembled based on the previously described data (26, 28, 29) as an initial estimate.

Simulations of each lipid were set up using a modified version of *insane.py* python script (30), embedding one copy of a PIM lipid in a 10 x 10 nm² phosphatidylcholine (PC) bilayer. The system was solvated with MARTINI water (26) and neutralized with 150 mM NaCl, followed by minimization using the steepest descents algorithm. The system was then simulated for 3 μ s using a timestep of 20 fs at 310 K. The lipids, solvent and ions were temperature coupled separately. The velocity rescale (31) and Parrinello-Rahman (32) coupling methods were used with the time constants $\tau_T = 1.0$ ps and $\tau_p = 12.0$ ps for temperature and pressure, respectively. Simulations were run using GROMACS version 2021.3 (33). The reaction-field algorithm (34) was used for electrostatics interactions with a cut-off of 1.1 nm. A single cut-off of 1.1 nm was used for the van der Waals interaction. Five repeats were performed for each lipid (**SI Figure 2**).

203 Generating atomistic data

Parameters for the atomistic (AT) Ac_xPIM_x lipids were generated using the CHARMM force field (35). The CHARMM-GUI (36) server was used to set up the lipid systems for GROMACS with the CHARMM36m force field (37, 38). One PIM lipid was embedded in an 8 x 8 nm² PC bilayer. The system was solvated with 150 mM NaCl and minimized and equilibrated as per the CHARMM-GUI Membrane Builder workflow (39). The system was further minimized using the steepest descents. Simulations were run for 2 µs using a timestep of 2 fs at 310 K. The lipids, solvent and ions were temperature coupled separately. The velocity rescale (31) and C-rescale coupling methods were used with the time constants τ_T = 0.1 ps and τ_p = 1.0 ps for temperature and pressure, respectively. Simulations were run using GROMACS version 2021.3 (33). The particle mesh Ewald (PME) (40) method was used for electrostatic interactions with a cut-off of 1.2 nm. A single cut-off of 1.2 nm was used for the van der Waals interaction. Three repeats were performed for each lipid (SI Figure 2).

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- 218 Refining coarse-grained parameters
- The AT and CG representations of the four main PIM lipids are shown in **Figure 1C**.
- 220 The distributions of the distances and angles were measured using the *gmx* tools
- 221 (distance, gangle and analyze). For the AT simulations, the atoms were grouped
- according to their mapping and its center of geometry was used for calculations. The
- values for each bond/constraint and angle were iteratively refined based on the
- 224 comparison of probability distribution, and the results are summarized in **SI Figures**
- 3-8. When there was agreement between the AT and CG data, the solvent accessible
- surface area was measured using the *gmx* sasa tool to verify that the models behaved
- the same (**SI Figure 8C**). Diffusion, shape of the lipids, clustering of ions, aggregation
- 228 and interaction with other lipids were also measured on a single lipid level (SI Methods
- and **SI Figures 9-19**). As typical for MARTINI (41), dihedral terms are not defined in
- the parameters.

- 232 Measuring area per lipid
- 233 Before setting up the complete bilayer system, the area per lipid was measured. The
- 234 simulations were assembled as described above for CG, using a homogenous
- 235 Ac_xPIM_x membrane (**SI Figure 2**). The area of XY-dimension was measured using
- 236 gmx energy and then dividing the area by the number of lipids in one leaflet over the

course of the trajectory, and the final values were extracted using *gmx analyze*, as per the protocol described on the MARTINI website (http://cgmartini.nl/index.php/tutorials-general-introduction-gmx5/bilayers-gmx5#Area-per-lipid).

- Bilayer composition and set-up
- The ratio of lipids in each bilayer was obtained by using their molecular weight and the previously reported dry masses (13) of the individual lipids, and the exact calculations can be seen in **SI Figure 20**. Since specific apolar lipids were not named, they were not included in this study. TMM was not included as there is no available refined AT model, the behavior of the mycolic acid is predicted to be complicated (42), and there is a relatively small proportion of TMM predicted to be present in the plasma membrane. The final composition of the plasma membrane is shown in **Figure 1A**.

The PI and PE lipids in mycobacteria are slightly different in structure to the corresponding *E. coli* lipids described in the MARTINI force field (26, 28): a methyl group replaces the alkene found in one of the acyl chains (**SI Figure 1**). Before assembling the membrane, the parameters for these lipids were modified by adapting the refined Ac₂PIM₂ acyl tail parameters and changing the existing tail. The CL parameters were transferred over from the MARTINI 3 beta force field (21).

Simulations of the bilayer were set up using a modified version of *insane.py* python script (30) using the composition shown in **Figure 1A**, where the area per lipid for the periplasmic membrane was set to 0.92 nm² and the cytoplasmic leaflet set to 1.13 nm². The systems were then treated the same as described for the initial CG systems with added equilibration steps as per the Membrane Builder workflow (39) before the production simulation. There were two types of system assembled to represent the asymmetric membrane previously mentioned (13). The first one had an initial simulation box size of 20 x 20 x 15 nm³ and one repeat was performed for 10 µs at 290 K, 300 K, 310 K, 320 K and 350 K (**Figure 2A**). The second one had an initial box size of 50 x 50 x 15 nm³ and a single repeat was performed for 10 µs at 310 K (**Figure 2C** and **SI Figure 2**). The number of lipids in each system is summarized in **SI Figure 21**.

To test how the membrane properties change with differing compositions, the same assembly procedure was followed to set up a symmetric membrane and an 'PIM-enriched' membrane (**SI Figure 22**), with one repeat for each in a simulation box of $20 \times 20 \times 15 \text{ nm}^3$, and another in a box of $50 \times 50 \times 15 \text{ nm}^3$, both at 310 K (**SI Figure**

2). Another membrane to capture an experimentally extracted membrane was simulated for 10 µs in a 20 x 20 x 20 nm³ simulation box at 310 K. The number of lipids in each system is summarized in SI Figure 21 and their compositions in SI Figure 22

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- Behavior of the membrane
- 276 To calculate the bending rigidities, the strategy proposed by Fowler et al. (43) to extract the bending modulus from CG simulations was followed. The membrane 278 midplane position in Monge representation was determined by extracting the 279 coordinates of all CG beads at the extremities of lipid fatty acid chains. Using a builtin function of the Mathematica software package, the positions were interpolated to get a smooth function before Fourier-transforming with a fast Fourier transform 282 algorithm. From the so-obtained Fourier modes, the spectral density (or power 283 spectrum) was estimated for a tensionless membrane:

$$S(q) = L^2 k_B T \left(1/(\kappa q^4) + 1/(\sigma_p r q^2) \right)$$

- where L^2 is the projected area in the (x,y) plane, k_BT is the thermal energy and $\sigma_n r \sim 0.1 \text{ J/m}^2$ is the tension associated with lipid protrusions at the nanometer scale. $S(q)q^4$ was plotted as a function of q and fit it with a second order polynomial P(q) = $a + bq^2$, from which estimates of κ (and $\sigma_n r$ if needed) were obtained (SI Figure 23). The error bars are standard deviations provided by the fitting function in Mathematica.
- The density of each constituent was measured using *gmx density* (**Figure 2B**). Diffusion, bilayer thickness and number of neighbors was measured using LiPyphilic (44) (Figures 2D,E, SI Methods and SI Figures 24-27). The xy-positions of single lipids were tracked with PLUMED (45) (Figure 2E). Area per lipid for the membrane was calculated using FATSLiM (46) (Figure 2F). Plots were created using Matplotlib (47).

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- 297 Analysis of the residue distribution in membrane proteins
 - All predicted protein structures from Mtb (taxonomy id:83332) and E. coli (taxonomy id:83333) labelled as transmembrane on UniProt (48) were downloaded from the AlphaFold database (49), totaling 729 for Mtb and 1,229 for E.coli. Alphafold was used to obtain a workable dataset of structures, expanding upon the number of experimentally determined *Mtb* and *E. coli* membrane protein structures available from the PDB. Each of 1,958 membrane protein structures was orientated in a membrane

using memembed (50) with the *-n in* flag, then converted to CG using martinize2 (51) and inserted into either the asymmetrical mycobacterial (see **Figure 1A**) or *E. coli* membrane (75% PE, 15% PG, 10% CL) using a modified version of the *insane.py* python script (30). Each system was simulated for 10 ns. The residues within 8 Å of the lipids were selected and the density over the z-coordinate plotted (**Figure 3, SI Figure 28**).

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Antibiotic simulations

Bedaquiline (BDQ) and isoniazid (ISZ) were mapped to CG using PyCGTOOL (52) following 200 ns AT simulations with parameters from CHARMM-GUI (53) (Figure **4A**). The pK_a of the amine in BDQ is 8.91 (54), suggesting it will be protonated at physiological pH. This was considered when assigning the MARTINI 3 bead type (Figure 4A). To validate the use of the protonated state, the same simulations with a neutral BDQ molecule were performed and analyzed (SI Figures 29A-C). This was done by replacing the Q1 particle with a SC3 bead type (Figure 4A). In addition, the LogP for these antibiotics were calculated by perturbing the antibiotic in a box containing 878 MARTINI waters or 363 octanol and 32 waters. The Lennard-Jones parameters for each molecule were decoupled over 11 evenly spaced λ windows, for 50 ns per window. Simulations were run with 20 fs time steps in the NPT ensemble with the V-rescale thermostat at 323 K and an isotropic Parrinello-Rahman pressure coupling. Free energies were then computed using gmx bar and converted to water/octanol partitioning constants. The values were then compared to data from the ALOGPS online server (55). Alongside this, gmx sasa was used to calculate the solvent accessible surface area for the antibiotics in both CG and AT. The calculations were performed with 4800 dots per sphere and a probe size of 0.4 nm. Modified van der Waals radii for CG beads of 0.264, 0.230, and 0.191 nm for R, S and T beads were used respectively. For the Connolly surface generation, the calculations used 240 dots and a probe of 0.4 nm (SI Figure 30).

BDQ was first simulated with the mycobacterial membrane alone to confirm association with the membrane (**SI Figures 2,31**). The *Mtb* a- and c-subunits of F-ATPase, BDQ's target in the mycobacterial membrane, was modelled using SwissModel (56) based on a structure from *Mycobacterium smegmatis* (25) (PDB: 7JGC, **Figure 4C** and **SI Figures 32A,B**). The sequence identity between *Mtb* and *M. smegmatis* for the F-ATPase was calculated using Clustal Omega (57) at 80% and

72% for c- and a-subunits, respectively. The system with the protein in the asymmetric mycobacterial membrane (see **Figure 1A**) was assembled using martinize2 (51), memembed (50) and the modified *insane.py* python script (30), with eight molecules of BDQ placed either in the periplasmic or cytoplasmic leaflet with five repeats in each membrane (**SI Figure 2**). Simulations were run for 10 µs using the same setting as described above. The xyz-positions of BDQ, the lipids and the backbone beads were tracked over the course of the simulations with PLUMED (45) and plotted with Matplotlib (47), and the results are shown in **Figure 4D** and **SI Figures 32C,D, 33**. The density of each constituent was measured using *gmx density* and plotted using Matplotlib (47) and the results are shown in **Figure 4E** (with representative positions shown in **Figure 4F**). The interaction of BDQ with the protein was calculated using PyLipID (58) (**Figure 4G** and **SI Figures 34,35**). The same simulation set up was performed with a model *E. coli* membrane (75% PE, 15% PG, 10% CL) and with the symmetric mycobacterial membrane (**SI Figure 22A**). The z-position of BDQ is shown in **SI Figure 32** and **SI Figures 36,37**.

PMF calculations

The potential of mean force (PMF) calculations for the F-ATPase-BDQ interaction were performed (59). First, a representative pose of BDQ bound to the protein was produced using PyLipID from equilibrium simulations and built into a 12 x 20 x 11 nm³ asymmetric mycobacterial membrane (see **Figure 1A**) membrane and minimized and equilibrated, as described above. Light (50 kJ/mol/nm²) *xy* positional restraints were added to Ala 66 on three c-subunits to prevent the protein from rotating in the membrane. Following 50 ns of equilibration, the BDQ was steered away from the protein along the *y* axis at a rate of 1 nm/ns with a 1,000 kJ/mol/nm² restraint potential. Frames were extracted at 0.1 nm spacing along this coordinate to seed a total of 58 x 1.5 µs production simulations with a static 1,000 kJ/mol/nm² umbrella potential imposed to keep the system in the same position along the reaction coordinate. The PMF mdp input files can be downloaded at https://github.com/chelsea-brown/PIM-lipids/tree/main/PMF-files. The PMF profiles were then constructed using the weighted histogram analysis method in GROMACS (*gmx wham* (60, 61)), and employing 200 rounds of Bayesian bootstrapping to report statistical accuracy (**Figure 4H**).

For the membrane crossing PMFs, BDQ or ISZ was placed free in the solvent phase, 7 nm away from the membrane periphery in either an asymmetric

mycobacterial membrane (see **Figure 1A**) or *E. coli* membrane. The drug was then steered towards and through the membrane and into the solvent phase on the other side. Windows were extracted, simulated (1 µs for ISZ per window and 2 µs per window for BDQ) and analyzed as described above (**Figure 4B** and **SI Figure 29D** for BDQ neutral). The -cycl option was imposed when running *gmx wham*.

Results

PIM properties

As an initial step towards developing an in-silico model of the mycobacterial plasma membrane, we first examined lipid properties in our CG and AT models. As shown in SI Figures 3-8, the probability distributions of bonds and angles from the CG and AT simulations align well, which illustrates that the behavior of the acyl chains is similar to other phospholipids (30). The Ac_xPIM_x aggregation in the AT and CG simulations is comparable and shows no permanent clustering. When lipids did come into contact, all areas of the molecule appear to play an equal role in the interactions (SI Figures **15,17**). The interactions were not dominated by the sugars, as was seen with the previous iteration of the MARTINI 2 force field (62), and the phospholipids present in the mycobacterial membrane did not have strong interactions with the PIM lipids (SI Figure 18,19). The PIM lipids show a higher affinity for ions than PC in CG simulations. but the effect is less apparent in AT simulations (SI Figure 14). While the ion concentrations were the same, the number of ions in the simulation box were different (an order of magnitude higher for CG), which could explain these results. Overall, CG simulations of these lipids behave similarly to AT models (interactions with ions as a minor exception), therefore opening the door to significantly longer simulations by decreasing the degrees of freedom in the system.

The area per lipid in CG for each species in a PC bilayer was found to be 0.93 nm² (AcPIM₂), 1.15 nm² (Ac₂PIM₂), 1.01 nm² (AcPIM₆), 1.14 nm² (Ac₂PIM₆), 0.6 nm² (PI), 0.55 nm² (PE) and 1.25 nm² (CL). The diffusion coefficients for each PIM species in a PC bilayer at 310 K are as follows: 6.7 x 10^{-7} cm²/s (AcPIM₂), 1.1 x 10^{-7} cm²/s (Ac₂PIM₂), 7.6 x 10^{-7} cm²/s (AcPIM₆), 6.5 x 10^{-7} cm²/s (Ac₂PIM₆) (**SI Figure 9**). The difference between AcPIM_x and Ac₂PIM_x shows the effect of the extra acyl tail in terms of how freely these lipids diffuse through the membrane.

The lipid shapes are approximately the same from AT to CG (SI Figures 10-13), which agrees with the comparison of the surface areas (SI Figure 8C) and the sugar-phosphate z-distances in Ac_xPIM₆ (SI Figures 5C,7C). The additional mannose moieties project upwards away from the membrane in both sets of simulations. Interestingly, the lipid tail region is measured to occupy approximately the same amount of space with three or four acyl chains. This is likely due to the placement of the fourth tail, projecting downwards from the inositol sugar (as highlighted in red in Figure 1B) and hence inhabiting space close to the other acyl chains.

Taken together, the lipid properties in our CG model broadly replicate those seen with AT resolution. This provided the confidence to assemble these lipids into a complex bilayer.

Mycobacterial membrane biophysical properties

Traditional MD simulations of bacterial membranes use symmetric bilayer representation. However, a study has shown that the mycobacterial plasma membrane could be an asymmetric bilayer, and we used the published composition of the mycobacterial plasma membrane (13) (Figure 1A) (excluding apolar lipids and TMM) to assemble an asymmetric bilayer (Figures 2A,C). This bilayer was found to be stable during 10 µs simulation (SI Movie 1), and in good agreement with the literature (8, 13). The thickness of the plasma membrane in our simulations was at 5.3 ± 0.1 nm (Figure 2B), which is slightly narrower than values obtained from imaging studies (6.3 nm for *Mycobacterium bovis* and 7 nm for *Mycobacterium smegmatis*) (63-65). However, we did observe thicknesses of up to 7 nm in the density plot (Figure 2B). *In vitro* culture conditions or growth stages of the bacteria (65) impact the thickness on the membrane and could explain any minor differences with our model.

To test the phase behavior of this membrane, we performed simulations at various temperatures ranging from 290 K to 350 K. For the whole range of temperatures, lipids were in the liquid phase and allowed to diffuse freely (**Figure 2E** and **SI Figure 24**). The calculated diffusion coefficients for each species in the periplasmic leaflet at 310 K were as follows: 9.9 x 10⁻⁸ cm²/s (AcPIM₂), 6.9 x 10⁻⁸ cm²/s (AcPIM₆), 1 x 10⁻⁷ cm²/s (Ac₂PIM₆), 1.3 x 10⁻⁷ cm²/s (PI), 1.5 x 10⁻⁷ cm²/s (PE) and 1.2 x 10⁻⁷ cm²/s (CL). In the cytoplasmic leaflet, the values were 6.1 x 10⁻⁸ cm²/s (AcPIM₂) and 8.1 x 10⁻⁸ cm²/s (Ac₂PIM₂). Compared to isolated PIMs in a PC membrane, the diffusion in this plasma membrane was roughly one order of magnitude slower.

AcPIM₂, which is the only lipid in both leaflets, diffused ~40% slower in the cytoplasmic membrane compared with the periplasmic leaflet. Using a previously reported model of mammalian plasma membrane (22), the diffusion coefficients of PE and PI were calculated to be 3.3×10^{-7} cm²/s and 2.8×10^{-7} cm²/s, showing that the diffusion in the mammalian membrane is equivalent to that in the mycobacterial plasma membrane. The membrane stiffness of our mycobacterial membrane was significantly lower at $\kappa = 8.2 \text{ k}_B\text{T}$ compared to $13.9 \text{ k}_B\text{T}$ and $19.1 \text{ k}_B\text{T}$ for the mammalian and PC membrane models, respectively (SI Figure 23), highlighting a unique dynamic behavior for the mycobacterial plasma membrane.

Lipid clustering was moderate and membrane composition remained heterogenous over the course of each 10 µs simulation (**Figure 2C**). This can also be seen in **Figure 2D** where the number of surrounding lipids of the same type for each species in the periplasmic membrane was roughly equivalent to that of any other lipid species. The distribution of the area per lipid in the membrane also suggests high heterogeneity (**Figure 2F**). The average area per lipid in the membrane over the course of the simulation was 0.89 nm² and 1.18 nm² for the periplasmic and cytoplasmic leaflets, respectively. Thus, the cytoplasmic leaflet appears to be a little denser than the periplasmic leaflet. This can be related to the packing of the four acyl chains of Ac₂PIM₂ lipids present in high concentration in this leaflet (66). A movie for the change in area per lipid for each leaflet over the course of the simulation can be found in **SI Movies 2,3**.

As was seen for the individual lipids in CG the overall bilayer attracted ions; both Cl⁻ and Na⁺ concentration was much higher close to the bilayer, especially around the sugar head groups compared to bulk solution (**Figure 2B**). It has been shown that lipid-ion interactions can affect the biophysical properties of the membrane, such as fluidity and stiffness, as well as the structure, which could modify the interaction with proteins (67).

The mycobacterial plasma membrane has been reported to change at different growth stages (7, 68), but the study that originally proposed the composition was based solely on late-exponential-phase cells. To explore a wider range of possible membrane compositions, we designed an 'PIM-enriched' bilayer, where the amount of PIM lipids in the periplasmic leaflet was increased to reflect cells in rapid growth (7) (see compositions in **SI Figure 22**). In this case, the overall biophysical properties of that leaflet were very similar to those seen in **Figure 2**, but interestingly a change in

the diffusion and density of the cytoplasmic leaflet was observed (**SI Figures 26,27**). Increasing the proportion of lipids in the periplasmic leaflet that have 3 or 4 acyl chains could mean that the lipid tails make more inter-leaflet interactions and hence slow down diffusion and make the leaflet denser. This decrease in fluidity in the membrane could impact antibiotic diffusion through the membrane (13, 69) which would make the bacteria less susceptible to treatment at a time when it is most vulnerable.

We have also tested alternative plasma membrane compositions with a symmetrical configuration. First, we performed a model of a membrane recapitulating the composition being experimentally derived using a combination of mass spectrometry and NMR (see composition in SI Figures 22,38). This symmetrical configuration is stable, and the dynamics of PIM lipids in CG-MD simulations agreed with solid state NMR results (see SI Figures 38,39). This data shows that the acyl chains extremities are highly mobile, while the core of the acyl chains displayed a more rigid structure. Interestingly, the sugar moieties adopted an intermediate dynamic regime, as seen from experimental NMR results (SI Figure 38E) and MD simulations (SI Figure 39B). We then designed a model of symmetrical membrane system with a lipid composition that mimics *Mtb* membranes (see composition in SI Figure 22A). This membrane showed similar interactions of lipids compared to the asymmetric composition and also similar rates of diffusion (SI Figures 25,27). Clustering of ions around the sugar headgroups was still observed.

Taken together our observations suggest that PIM lipids play a critical role in the dynamic and structuration of mycobacterial plasma membrane. Both the asymmetric and symmetric configurations of the membrane were equally stable and our analyses show that both share similar dynamic properties. This work demonstrates the plausible arrangements of the plasma membrane and provides a mechanism to easily study new membrane compositions based on future experimental results.

Asymmetrical distribution of residues at the surface of Mtb membrane proteins. The lipids of the plasma membrane do not exist in isolation, but rather in cooperation with membrane proteins. These proteins could provide insights into how this membrane exists natively (70). To explore this, we used the accurate, fast and reliable workflows of AlphaFold (49) and memembed (50), to insert all transmembrane proteins from Mtb and E. coli into their native-like membranes (Figure 3A,B). We then assessed the distribution of residues in proximity to the lipid bilayer (Figure 3C).

For *E. coli*, the distribution is as expected and similar to results previously shown (21). The asymmetry of Arg and Lys reflect the 'positive inside rule' (71-73) present in bacteria which aids in the correct insertion of the transmembrane helices. The symmetric nature of the lipid bilayer is otherwise reflected in the distribution of the residues (**Figure 3C**, **SI Figure 28**). The *Mtb* distribution of Arg also shows a strong 'positive inside rule'. This is at a higher intensity than in *E. coli*, likely due to the low levels of Lys observed (**Figure 3C**). Furthermore, the decrease in the bulky Phe and Tyr residues (**Figure 3C**) in the cytoplasmic leaflet compared to the periplasmic leaflet agrees with the increased density seen in an asymmetric membrane (**Figure 2B** and **SI Figure 26B**).

These results add evidence to an asymmetric model previously proposed (13). In addition to validating the model proposed in this work, it could aid understanding of how mycobacterial membrane proteins differ from those found in other organisms and hence aid targeted therapies.

Asymmetric interaction of the membrane with antibiotics

We were interested in testing how the organization of this membrane affects the behavior of other molecules, such as antibiotics. We tested two widely used anti-TB antibiotics: Isoniazid (ISZ) and Bedaquiline (BDQ) (**Figure 4A**). To validate the behavior of these small molecules as well as the bond lengths/angles being account for through PyCGTOOL (52) (**SI Figure 30**), the LogP and solvent accessible surface area were calculated for each and compared to predicted values and atomistic values, respectively. The values for LogP for neutral BDQ (5.92 for CG, 6.37 predicted) and ISZ (-1.10 for CG, -0.71 predicted) agree very well, as do the solvent accessible surface area values (BDQ values 13.01 ± 0.49 nm² for CG and 13.69 ± 1.11 for AT; ISZ values 7.48 ± 0.14 nm² for CG and 7.35 ± 0.04 for AT).

ISZ is a first line anti-TB treatment that targets InhA, a cytosolic enzyme that is essential for mycolic acid synthesis (74). BDQ on the other hand is a last line antibiotic that targets the membrane protein complex of the ATP synthase (75, 76). The PMF results for ISZ and BDQ show that the mycobacterial membrane behaves as expected regarding the passage of a small molecule through a lipid rich membrane, showing favorable interactions with the largely hydrophobic BDQ and unfavourable interactions with hydrophilic ISZ (**Figure 4B**). The positive charge present on BDQ could account for the sharp free energy barrier at the membrane midplane. To assess the effect of

the charge on the passage of BDQ since the protonation state is unknown, we modeled a neutral BDQ where the charged Q1 bead (see Figure 4A) was replaced by SC3. The PMF results with the neutral BDQ show a reduction in the energy barrier at this point (SI Figure 29D). We note that the passage of these drugs through an E. coli inner membrane is symmetric from the mid-plane of the membrane as expected. The interactions of BDQ with the periplasmic leaflet of the mycobacterial membrane were ~6 kJ/mol stronger than for the simple membrane and the mycobacterial cytoplasmic leaflet. This is possibly due to the presence of tetraacylated PIMs in the mycobacterial periplasmic leaflet. Roughly the same difference between the mycobacterial and simple membranes was observed for ISZ at the mid-membrane region, showing less favorable passage through the membrane. These results suggest that ISZ, a first-line TB drug, is unlikely to freely diffuse through the membrane, which is in contrast with the previously proposed uptake mechanism based on passive diffusion (77-79). Small differences were seen for ISZ interactions at the interfaces of the mycobacterial and simple membrane periplasmic membrane. This could be due to the presence of certain PIM lipids in this leaflet. In our simulations, when the BDQ starting position was in the bulk solvent, the drug quickly associated with the membrane, as shown in SI Figure 31, and displayed no strong preference for either leaflet in these simulations.

BDQ has been shown to target the mycobacterial ATP synthase (Rv1304-Rv1311), a membrane protein complex. A cryo-EM structure from *M. smegmatis* (PDB: 7JGC) shows multiple binding sites for BDQ at the interface of the c-subunits and the interface between the a- and c- subunits (25), where a highly negatively charged area is located (SI Figure 32A,B). Using the available cryo-EM data, we performed 10 simulations of 10 µs each to examine how *Mtb* ATP synthase complex behaves in our asymmetric mycobacterial plasma membrane model. ATP synthase was stable in the membrane and that there were no significant perturbations of the bilayer by the protein. The lipids of the mycobacterial membrane did not show any strong interactions with the protein (Figure 4D and SI Figure 33), apart from CL that localizes in the a-subunit and around the c-ring (Figure 4D) in positions similar to those observed in a previous study (80). The exact values for CL occupancy of each residue are shown in SI Figure 35A. Minimal interactions between BDQ and the ATPase occurred when the drug started in the periplasmic leaflet; however, when BDQ was introduced in the cytoplasmic leaflet, significant occupation of the binding sites on

ATPase were observed (Figures 4D-G and SI Figures 32C,D). The interactions were reduced with the neutral BDQ model (SI Figures 29A-C). Of note, no such leaflet differences were observed when we tested the *E. coli* symmetric membrane model (SI Figures 36C,D; SI Figure 37), but with a symmetric mycobacterial membrane (SI Figure 32). This suggests that the lipid composition plays a crucial role in the regulation of antibiotic interactions.

Furthermore, the interactions of the charged BDQ we observed matched well with what was previously seen in the cryo-EM structure (25), with an occupancy of ~30% of the simulation time on some subunits (**Figure 4G**). The occupancy value averaged over all subunits is shown in **SI Figure 35B**. The simulations were able to identify all three types of sites reported previously (25). Most interactions were seen through the leading site (46% of the time, $K_{off} = 5.2 \, \mu s^{-1}$) followed by the lagging site (22% of the time, $K_{off} = 7.6 \, \mu s^{-1}$) and finally further interactions around the rotor (36% of the time, $K_{off} = 1.2 \, \mu s^{-1}$, **SI Figure 34**). For the leading site, the PMF calculations confirm this as a binding site, giving a moderate energy well of approximately 7 kJ/mol (**Figure 4H**) and making it equivalent to a typical cholesterol-protein binding interaction (81). Tracking the z-position of the central bead from the antibiotic over the course of the unbiased simulations with the protein shows that the binding sites are occupied to some extent by each of the drug molecules and some are flipped to the outer leaflet from the plasma leaflet, as shown in **SI Figure 32C**.

These studies illustrate that our asymmetrical mycobacterial plasma membrane model can be combined with studies of antibiotics, as well as with studies of membrane proteins and their interactions with lipids and/or antibiotics. Taken together, our simulations suggest that BDQ must enter the target protein from the cytoplasmic side, which agrees with resistance mechanisms against BDQ that involve upregulation of MmpL5, a drug efflux pump (82, 83). Importantly, an *E. coli* membrane could not capture this feature, highlighting that the mycobacterial membrane model is a more accurate and functionally relevant representation of the *Mtb* plasma membrane.

Discussion

Here, we provide models for lipids constituting the mycobacterial plasma membrane focused on the PIM lipids, as they represent more than 50% of the total lipid content in *Mtb*. Thus, understanding how these lipids behave on an individual level and as a

constituent of a membrane could provide key insight into the intrinsic resistance of *Mtb* to antibiotics.

In this CG model, clustering of ions around these lipids was observed at both the single lipid and bilayer level, probing how this affects the biophysical properties of the bilayer and whether this could be exploited for treatment of TB is an interesting area for future research. The lipids did not cluster together excessively over the timescales studied and all diffused well through the membrane. This confirmed that the mycobacterial cell envelope is dynamic, which could potentially be an important insight into how this cell wall functions. A low membrane bending rigidity compared to a PC and eukaryotic plasma membrane is interesting and could suggest the importance of other cell envelope components in maintaining the shape of the cell.

The simulations confirmed that our asymmetric plasma membrane model is stable with a composition of over 50% PIM lipids. However, considering that this arrangement has not been experimentally validated, a symmetric and 'enriched PIM' asymmetric membrane was also shown to share similar biophysical properties. The high proportion of PIM lipids resulted in some unique properties, most notably lower membrane stiffness, which may have important implications for membrane microdomain formation, drug penetration, and integral membrane protein behavior. In this context, the integral membrane proteins from Mtb are of interest for the development of new antibiotics for TB (9), and we show evidence that our membrane model can be used to simulate proteins in a native lipid environment to determine any key lipid interactions, as illustrated by our study of antibiotic BDQ and its target, ATPase. In addition, an overall comparison of plasma membrane protein residues found in contact with the lipid bilayer revealed differences between Mtb and E. coli, supporting the asymmetric nature of the mycobacterial plasma membrane model. This potentially hints at topological differences between proteins in mycobacteria and those within other bacteria. If this is the case, currently unknown proteins which maintain the asymmetry might be present (such as those identified in animal cells (84)), and would constitute interesting drug targets to explore.

Potential limitations of our model stem from the reduced type of lipids we included in this initial development. Thus, future studies will focus on expanding the model to include apolar lipids (such as triglycerides (17)), TMM and LM/LAM. Moreover, future simulations with proteins and probing protein/lipid interactions with atomistic resolution could help further elucidate the role of these complex lipids.

In conclusion, our study is a starting point for building an entire mycobacterial cell envelope, as done for gram-negative bacteria (85). Other mycobacterial lipids have already been parameterized (86), and combining them with the model described here has the potential to significantly advance the field that has been lacking experimental strategies for plasma membrane investigations. Additionally, the mycobacterial cell envelope is known to change at different growth stages during its lifecycle (7, 68, 87, 88) and in response to different environments (69). With our model and set-up methods, as shown with the multiple compositions simulated, it will be possible to understand drug permeability or protein behavior at different stages of infection. Taken together, integrating our mycobacterial plasma membrane model into broader studies of this pathogen, its pathogen/host interactions, as well as into anti-TB drug discovery and development has a potential to reveal new functional insights and yield improved therapies.

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- Data availability: https://github.com/pstansfeld/PIM-lipids
- 658 Atomistic systems and CG-membrane set-up can be performed using respectively
- 659 CHARMM-GUI bilayer builder (https://charmm-gui.org/input/membrane.bilayer) and
- 660 CHARMM-GUI MARTINI bilayer Maker (
- 661 https://charmm-gui.org/?doc=input/martini.bilayer)

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Author contributions

- 664 C.M.B., R.A.C., W.I., P.J.S. and M.C. designed research; C.M.B., R.A.C., Y.G., Y.K.C.
- and E.L. performed computational research; C.M.B., R.A.C. and N.D. analyzed
- computational data; A.G., M.G., J.N. and A.L. performed experimental research and
- analysis; C.M.B., R.A.C., A.L., P.J.S. and M.C. wrote the paper; J.N., A.L., E.F., W.I.,
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691 Abbreviations

692 AT: Atomistic

693 ATP: Adenosine triphosphate

694 BDQ: Bedaquiline

695 CG: Coarse grained

696 CL: Cardiolipin

697 ISZ: Isoniazid

698 LAM: Lipoarabinomannan

699 LM: Lipomannan

700 MD: Molecular dynamics

701 Mtb: Mycobacterium tuberculosis

702 NMR: Nuclear magnetic resonance

703 PC: Phosphatidylcholine

704 PE: Phosphatidylethanolamine

705 PG: Phosphatidylglycerol

706 PI: Phosphatidylinositol

707 PIMs: Phosphatidyl-myo-inositol mannosides

708 PIPs: Phosphatidylinositol phosphates

709 PME: Particle mesh Ewald

710 PMF: Potential of mean force

711 TB: Tuberculosis

712 TMM: Trehalose monomycolate

713 VdW: van der Waal

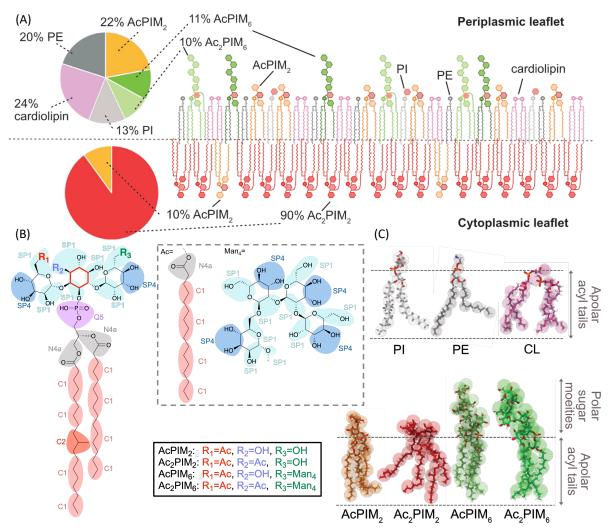


Figure 1: The structure of the mycobacterial lipids. (**A**) Schematic of an asymmetrical model of the mycobacterial plasma membrane and composition as previously defined (13). (**B**) Schematic of the core of the PIM lipids found in mycobacteria with the groupings for coarsegrained (CG) beads. The inositol core is highlighted in red. The bead types for MARTINI 3 are shown. (**C**) Overlay of the AT (sticks) and CG (spheres) models for each lipid, with chemical characteristics shown to the right.

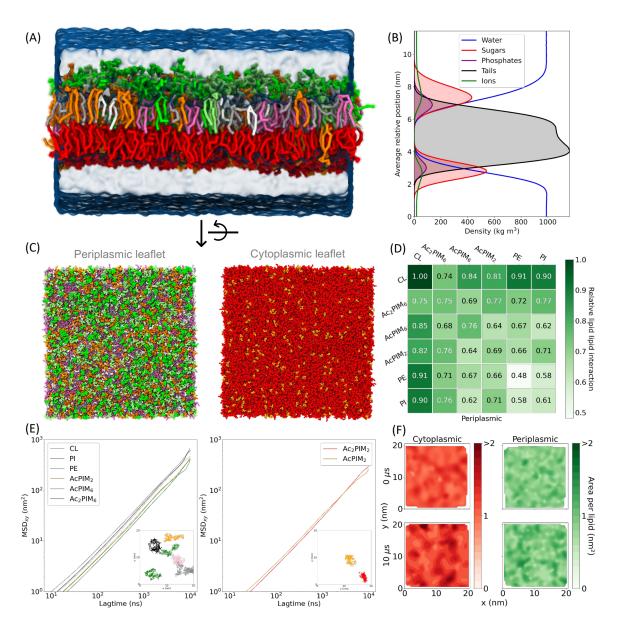


Figure 2: A mycobacterial membrane model. (**A**) Side view of the membrane with each lipid type depicted in a different color as shown in **Figure 1A**. (**B**) A density plot showing the density of water, sugar groups, phosphate groups, tail groups and ions over the simulation box. (**C**) Snapshots of the periplasmic and cytoplasmic leaflets whose compositions are: $AcPIM_2 22\%$, $AcPIM_6 11\%$, $Ac_2PIM_6 10\%$, CL 24%, PE 20% and PI 13% (periplasmic leaflet) and $AcPIM_2 10\%$ and $Ac_2PIM_2 90\%$ (cytoplasmic leaflet). The system size is $50 \times 50 \times 15$ nm. (**D**) Relative number of neighbors of each lipid type for the periplasmic membrane. (**E**) Mean squared displacement (MSD_{xy}) (nm^2) as a function of lagtime (ns) for each lipid type in the periplasmic and cytoplasmic leaflets. The inserts show the position of the phosphate group of each lipid type over the last 500 ns of the simulation. (**F**) Contour plots of the area per lipid (nm^2) in each leaflet at the starting frame (upper) and final frame (lower) of the simulation. A darker color indicates a larger area per lipid.

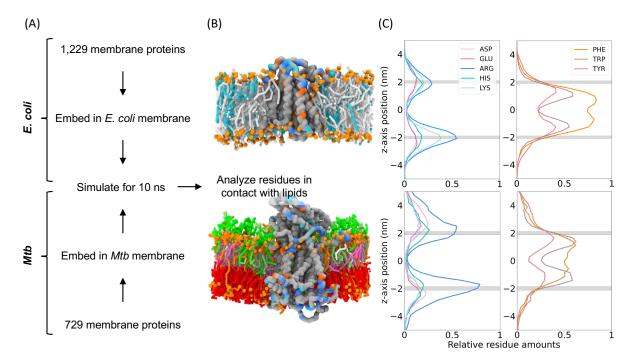


Figure 3: The asymmetry of the mycobacterial membrane. (**A**) A simplified workflow for the lipid contacts analysis from all membrane proteins. (**B**) An example of a protein (*E. coli* protein MraY UniProt id: P0A6W3 and *Mtb* transporter MmpL3 UniProt id: P9WJV5) embedded in either an *E. coli* or mycobacterial membrane. The orange spheres represent phosphates. For *E. coli* membrane the cyan sticks show PG and the grey sticks PE. For mycobacterial membrane, the lipid sticks are shown in the colors illustrated in **Figure 1A**. Selected residues are colored according to the key in (C). (**C**) Graphs for *E. coli* and *Mtb* (top and bottom respectively) showing the relative abundance of selected residues within 8 Å of lipids. The gray lines show the position of the phosphates. The cytoplasmic region is shown by negative z values, the periplasmic region by positive values.

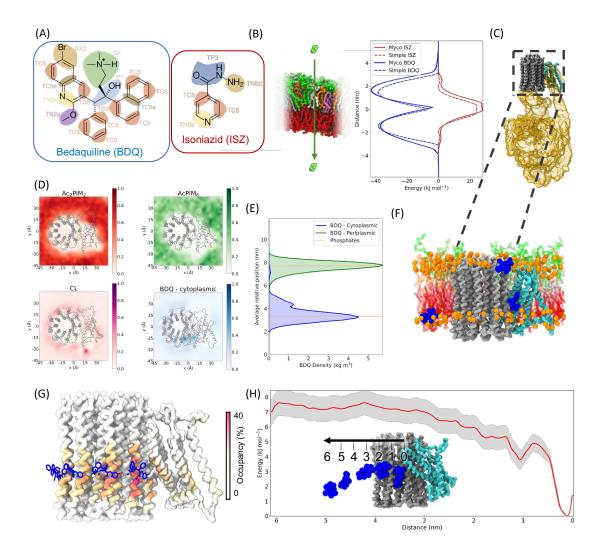


Figure 4: The behavior of the antibiotics with mycobacterial membrane and proteins. (A) Chemical structures of BDQ and ISZ with the CG groupings overlaid and bead types shown. (B) PMFs of the two antibiotics being pulled through either a mycobacterial or simple membrane in the z-direction. BDQ is shown in blue and ISZ in red, with the mycobacterial membrane results having a solid line and simple membrane having a dashed line. The error is shown in grey. A schematic of the PMF is shown to the left. (C) Structure of Mycobacterium smegmatis ATP synthase (PDB: 7JG5) with the c-subunits shown in grey, the a-subunit shown in cyan, and the other components shown as a gold surface. (D) Density in the x and y dimensions of selected lipids and BDQ when starting in the cytoplasmic leaflet relative to the protein shown in grey. (E) Density of the phosphates (orange) and BDQ over the course of the simulations where the antibiotic started in either the periplasmic leaflet (green) or the cytoplasmic leaflet (blue). (F) Snapshot of a single simulation containing a Mtb ATPase model and 8 x BDQ models showing the main positions BDQ occupied. Phosphates are shown in orange, BDQ shown in blue, c-subunits are shown in grey, and the a-subunit is shown in cyan. The lipid sticks are shown in the colors illustrated in Figure 1A. (G) Comparison of the highest occupancy sites identified with PyLipID (surface) and BDQ from the Cryo-EM structure (PDB: 7JGC) (sticks). (H) PMF of BDQ moving through a mycobacterial plasma membrane with the error shown in grey. A schematic of the PMF is shown as an insert.

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