

Matti Javanainen has made the text modifications to the manuscript in excellent way. He modified directly the manuscript tex file available in the GitHub and then made a pull request to add a new version of the manuscript into the GitHub repository. In addition, he used latexdiff which can be generated to show the pdf with changes.

After Matti had added his version I looked the changes through and agreed. Then I changed his version to the up-to-date version and made some minor modifications.

All the mentioned files can be found from the GitHub repository <https://github.com/NMRLipids/nmrlipids.blogspot.fi>
up-to date manuscript HGmodel_draft3.tex

Version Matti uploaded HGmodel_draft3_matti.tex

pdf showing the changes made by matti: HGmodel_draft3_matti_latexdiff.pdf

I think that this is a very good and convenient way to work collaboratively with the manuscript. If you want you can also make a pull request directly using the up-to-date manuscript file name (HGmodel_draft3.tex). Then I will accept the pull request, or we will do some discussion before acception. Using latexdiff is very good, but the history will also be saved automatically to the GitHub.

If you find this difficult (as it may be if you are not familiar with GitHub) you can also send your comments and modifications in some other way. However, keep in mind that it is very important to clearly report somehow what you have done. This is to ease my work to handle the input from large amount of contributors.

Towards atomistic resolution structure of phosphatidylcholine glycerol backbone and choline headgroup at different ambient conditions

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Phospholipids are essential building blocks of biological membranes. Despite of vast amount of accurate experimental data, the atomistic resolution structures sampled by the glycerol backbone and choline headgroup in phosphatidylcholine bilayers are not known. Atomistic resolution molecular dynamics simulations model would automatically resolve the structures giving an interpretation of experimental results, if the model reproduced the experimental data. In the present work we simulate phosphatidylcholine (PC) lipid bilayers with 13 different atomistic models, and we compare simulations with experiments (in fully hydrated conditions) in terms of C–H bond vector order parameters for the glycerol backbone and choline headgroups. None of the current models is not accurately enough to resolve the structure with experimental accuracy. However, closer inspection of three best performing models (CHARMM36, GAFFlipid and MacRog) suggest that improvements in the sampled dihedral angle distributions would potentially lead to the model which would resolve the structure. Despite of the inaccuracy in the fully hydrated structures, the response to dehydration, i.e. the tilting of the P–N vector more parallel to membrane normal, is qualitatively correct in all models. The CHARMM36 and MacRog models describe the interactions between lipids and cholesterol better than the Berger/Höltje model. **This work has been done as an open collaboration by using the `nmrlipids.blogspot.fi` as an communication platform. All the scientific contributions have been done through the blog and are publicly available. In addition, almost all simulation trajectories and files are made available in the Zenodo community <https://zenodo.org/collection/user-nmrlipids> which has become the most extensive publicly available collection of molecular dynamics simulation trajectories of lipid bilayers.**

I. INTRODUCTION

Phospholipids containing various polar headgroups and acyl chains are essential building blocks of biological membranes. Lamellar phospholipid bilayer structures have been widely studied with various experimental and theoretical techniques as a simple model for cellular membranes [1–8]. Phospholipid molecules are composed of hydrophobic acyl chains which are connected by a glycerol backbone to a hydrophilic

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headgroup. See Fig. 1 for the structure of 1-palmitoyl-2-oleoylphosphatidylcholine (POPC). The behaviour of the acyl chains in a lipid bilayer is relatively well understood [1–5, 8, 9]. The conformations sampled by the glycerol backbone and choline in a fluid bilayer are, however, not fully resolved since even the most accurate scattering and Nuclear Magnetic Resonance (NMR) techniques give only a set of values that the structure has to fulfil, but there is no unique way to derive the actual structure from them [9–18]. Some structural details have been extracted from crystal structure, ^1H NMR studies and Raman spectroscopy [19–25] but general consensus about the structures sampled in the fluid state have not been reached [9–18, 24, 25]. On the other hand, the glycerol backbone structures are similar for various biologically relevant lipid species (phosphatidylcholine (PC), phosphatidylethanolamine (PE) and phosphatidylglycerol (PG)) in various environments [26] and the headgroup choline structures are similar in model membranes and real cells (mouse fibroblast L-M cell) [27]. Thus, the solution of phosphatidylcholine glycerol backbone and choline structures would be useful for understanding a wide range of different biological membranes.

Classical atomistic molecular dynamics simulations have been widely used to study lipid bilayers [2–7]. As these models provide an atomistic resolution description of the whole lipid molecule, they have potential to solve the glycerol backbone and headgroup structures. In particular, the experimental C–H bond order parameters for the glycerol backbone (g_1 , g_2 and g_3) and choline (α and β) segments (see Fig. 1 for definitions) are among the main parameters used in attempts to derive lipid structures from experimental data [10–13, 15, 16, 18]. These parameters are also routinely compared between experiments and simulations for the acyl chains [2–6]. Thus, the structures sampled in a simulation model that reproduces these and other experimental parameters, automatically give an interpretation of the experiments. In other words they can be considered as reasonable atomistic resolution descriptions of the behavior of lipid molecules in a bilayer.

The glycerol backbone and choline headgroup order parameters have been compared between simulations and experiments in some studies [28–37], however much less frequently than for the acyl chains [2–6]. The main reason is probably that the existing experimental data for the glycerol backbone and choline headgroups is scattered over many publications and published in a format that is difficult to understand without some NMR expertise. In addition to the order parameters, also dihedral angles for glycerol backbone and headgroup estimated from experiments have been sometimes used to assess the quality of a simulation model [28, 38–42].

In this work we first review the most relevant experimental data for the glycerol backbone and choline headgroup order parameters in a phosphatidylcholine lipid bilayer. Then the available atomistic resolution lipid models are carefully compared to the experimental data. The comparison reveals that the CHARMM36 [31], GAFFlipid [33] and MacRog models [37] have the most realistic glycerol backbone and choline structures. We also compare the glycerol backbone and choline structures between the most often used (Berger

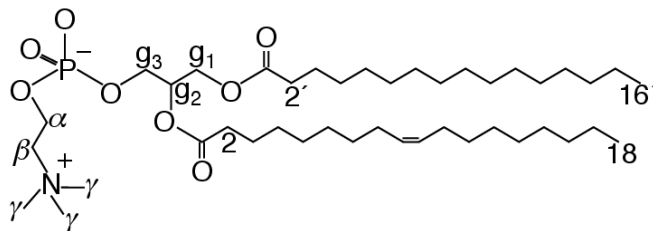


FIG. 1: Chemical structure of 1-palmitoyl-2-oleoylphosphatidylcholine (POPC).

based) lipid model [43] 1.If there are no objections, we will leave the naming convention of different Berger versions as it is now. and the best performing models, to demonstrate that by using the order parameters we can distinguish the more reasonable structures from the less reasonable ones. However, none of the current models is accurate enough to properly resolve the atomistic resolution structures.

In addition to fully hydrated single component lipid bilayers, the glycerol backbone and choline order parameters have been measured under a large number of different conditions. For example, as a function of hydration level [44–46], cholesterol content [35, 47] ion concentration [48–52], temperature [53], charged lipid content [51, 52], charged surfactant content [54], drug molecule concentration [30, 55, 56], and protein content [57, 58] (listing only the publications most relevant for this work and the pioneering studies). Awareness of the existence of this type of data allows the comparison of structural responses to varying conditions between simulations and experiments, which can be used to validate the simulation models and to interpret the original experiments. In this publication we demonstrate the power of this approach for understanding the behaviour of a bilayer as a function of hydration level and cholesterol concentration. Choline headgroup order parameters as function of ion concentration, and their relation to the ion binding affinity, are discussed elsewhere [59].

II. METHODS

A. Open collaboration

2.Should we write more about this?

This work has been done as an open collaboration by using `nmrlipids.blogspot.fi` as a communication platform. The approach is inspired by the Polymath project [60], however there are some essential differences. **The present**

approach was started by publishing a manuscript [61] discussing the glycerol backbone and choline structures in a Berger based model (the most used molecular dynamics simulation model for lipid bilayers). After publishing the initial manuscript, the open invitation for further contributions and discussion through the blog was presented. All the scientific contributions are done publicly through the blog and all contributors were offered coauthorship. The acceptance of authorship is based on the self-assessment of the authors' scientific contribution to the project. The contributions from each author are summarized in ?? (the location depends on the journal probably).

Almost all simulation data, including running parameters for reproduction and trajectories for further analysis, are collected to the nmrlipids community in Zenodo file repository <https://zenodo.org/collection/user-nmrlipids>. Thus, besides the main topic of this manuscript we present the most extensive available collection of simulations trajectories for lipid bilayers, which opens up numerous possibilities for different analyses with much less effort than previously required. Further information, e.g. scripts, figures and manuscript text files, are available in the GitHub organization <https://github.com/NMRLipids>.

B. Order parameters from experiments

3.DONE The order parameter of a hydrocarbon C–H vector is defined as

$$S_{CH} = \frac{1}{2} \langle 3 \cos^2 \theta - 1 \rangle, \quad (1)$$

where the angle brackets denote an ensemble average over the sampled conformations, and θ is the angle between the C–H bond and the membrane normal. The absolute values of order parameters can be measured by detecting quadrupolar splitting with ^2H NMR [62] or by detecting dipolar splitting with ^{13}C NMR [35, 63–65]. The measurements are based on different physical interactions and also the connection between order parameters and quadrupolar or dipolar splitting are different. The absolute values of order parameters from the measured quadrupolar splitting $\Delta\nu_Q$ (^2H NMR) are calculated using the equation $|S_{CD}| = \frac{4}{3} \frac{e^2 q Q}{h} \Delta\nu_Q$, where the value for the static quadrupole splitting constant is estimated from various experiments to be 170 kHz leading to a numerical relation $|S_{CD}| = 0.00784 \times \Delta\nu_Q$ [62]. The absolute values of order parameters from the measured dipolar splitting d_{CH} (^{13}C NMR) are calculated using equation $|S_{CH}| = \frac{4\pi \langle r_{CH}^3 \rangle}{h \mu_0 \gamma_h \gamma_c} d_{CH}$, where values between 20.2–22.7 kHz are used for $\frac{4\pi \langle r_{CH}^3 \rangle}{h \mu_0 \gamma_h \gamma_c}$, depending on the original authors [35, 63–65]. It is important to note that the order parameters measured with different techniques based on different physical interactions are in good agreement with each other (see Results and Discussion), indicating very high quantitative accuracy of the measurements. For a more detailed discussion see <http://nmrlipids.blogspot.fi/2014/02/>

accuracy-of-order-parameter-measurements.html 4.This is temporarily linked to the blog. Currently I think that the posts (with comments included) to which we want to cite are put into the Figshare in pdf format as they are in when the publication is submitted. Figshare allows commenting to continue the discussion. COMMENT: There is also the option to put the most relevant blog posts as a supplementary material. The advantage of this would be that the posts would be peer reviewed as well.

The absolute values of order parameters are accessible with both ^2H NMR and ^{13}C NMR techniques. However, only ^{13}C NMR techniques allow also the measurement of the sign of the order parameter [16, 63, 64]. The measured sign is negative for almost all the carbons discussed in this work as only α is positive [16, 63, 64]. For more detailed discussion about the sign measurements of the order parameters see <http://nmrlipids.blogspot.fi/2014/04/on-signs-of-order-parameters.html>

For most CH_2 segments in fluid phospholipid bilayer the order parameters are equal for both hydrogens attached to the same carbon. However, in some cases (e.g. g_1 , g_3 and C_2 carbon in the *sn*-2 chain) the order parameters are not equal and this can be observed with both ^2H NMR and ^{13}C NMR techniques. In the present work we call the phenomena of unequal order parameters for hydrogens attached to the same carbon as *forking* to avoid confusion with dipolar and quadrupolar splitting in NMR terminology. Forking has been studied in detail with ^2H NMR techniques by deuterating the R or S position in CH_2 segment, and the studies show that the forking arises from differently sampled orientations of the two C–H bonds, not from two separate populations of lipid conformations [26, 66].

C. Order parameters from simulations

The order parameters from simulations were calculated directly using the definition as in Eq. 1. For the united atom models the hydrogen positions were generated in the trajectories post-simulationally using the positions of the heavy atoms and the known hydrocarbon geometries. **For the statistical error estimates, the time average of order parameters were first calculated separately for each lipid in the system. Then it was assumed that different lipids are statistically independent entities (which should be the case in fluid phase) and the error of the mean for the average over individual lipids in the system was calculated and used as error bar for the order parameters.**

It has been recently pointed out that the sampling of individual dihedral angles might be very slow compared to the typical simulation timescales [67]. On the other hand, another recent study shows that the slowest rotational correlation functions of a C–H bond (g_1) reaches a plateau (S_{CH}^2) after 200 ns in the Berger-POPC-07 [68] model, and that the dynamics of this segment is significantly too slow in simulations compared to the experiments [69]. In practise, less than 200 ns of simulation data is enough for the order parameter calculation due to the average over different lipid molecules. In conclusion, if the sampling with typical simulation times is not enough for

the convergence of the order parameters, then the simulation models has significantly too slow dynamics.

D. Simulated systems

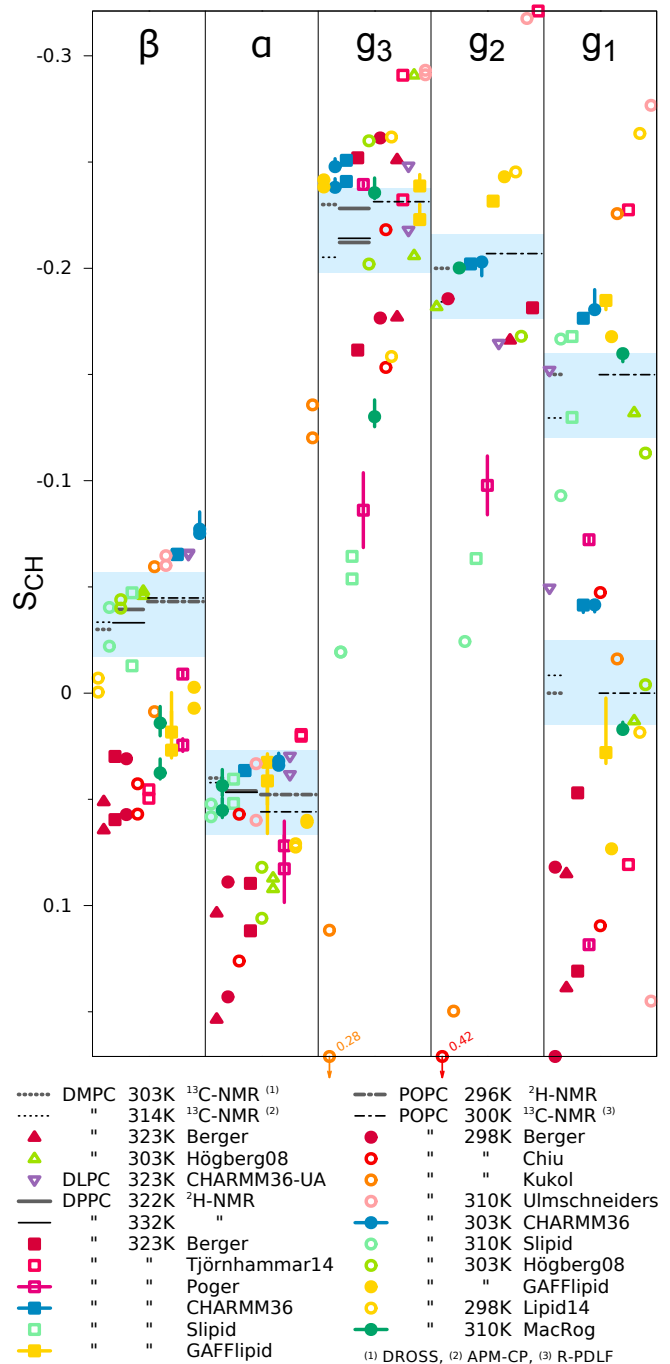
All simulations are ran with a standard setup for a planar lipid bilayer in zero tension and constant temperature with periodic boundary conditions in all directions by using the GRO-MACS software package [70] (version numbers 4.5.X–4.6.X) or LAMMPS [71]. The number of molecules, simulation temperatures and the length of simulations of all the simulated systems are listed in Tables I, II and III. Full simulation details are given in the Supplementary Information (SI) or in the original publications in case the data is used previously therein. Additionally, the files related to the simulations and the resulting trajectories are publicly available for some systems. The references pointing to simulation details and files are also listed in Tables I, II and III.

III. RESULTS AND DISCUSSION

A. Full hydration: Experimental order parameters for the glycerol backbone and headgroup

The specific deuteration of α -, β - and g_3 - segments of DPPC has been successful, allowing the absolute value order parameter measurements for these segments by ^2H -NMR [47–49, 53]. In addition, the absolute values of order parameters for all glycerol backbone and choline headgroup segments in egg yolk lecithin [63], DMPC [16, 64, 65], DOPC [120] and POPC [35, 120] have been measured with several different implementations of ^{13}C NMR experiments. In addition, the signs of order parameters in some systems are measured with ^{13}C NMR techniques [16, 63, 64]. The experimental values of glycerol backbone and choline order parameters from various publications [35, 49, 53, 64, 65] with the signs measured in [16, 63, 64] are shown in Fig. 2.

In general there is a good agreement between the order parameters measured with different experimental NMR techniques: Almost all the reported values are within a variation of ± 0.02 (which is also the error estimate given by Gross et al. [64]) for all fully hydrated PC bilayer, regardless of the variation in their acyl chain composition and temperature. Exceptions are the somewhat lower order parameters sometimes reported from measurements using ^{13}C -NMR [16, 63, 120]. These experiments are not shown in Fig. 2 as the reported error bars are either relatively large [16, 63], or the spectral resolution is quite low and the numerical lineshape simulations have not been used in the analysis [120]. Due to this end, it is highly likely that these reported lower order parameters are due to lower experimental accuracy and therefore we exclude them from our discussion. Motivated by the high experimental repeatability, we have highlighted in Fig. 2 the subjective sweet spots (light blue areas), within which we expect the



17. Markus Miettinen suggested that we should consider making one more figure where only experimental data would be shown and that would be discussed in Section III A. COMMENT: I do not think that we need such figure anymore since this figure is more clear now.

18. I think that this figure is now very good. There is also the interactive version by Hubert Santuz now in [https://plot.ly/ HubertSantuz/72/lipid-force-field-comparison/](https://plot.ly/HubertSantuz/72/lipid-force-field-comparison/) we should figure out which is the most practical way to put that behind permalink once it finalized (Zenodo, figshare or something else?) and then put a citation in the paper.

FIG. 2: Order parameters from simulations listed in Table I and experiments for glycerol and choline groups. The experimental values were taken from the following publications: DMPC 303 K from [64], DMPC 314 K from [65], DPPC 322 K from [53], DPPC 323 K from [49], POPC 296 K from [44], and POPC 300 K from [35]. The vertical bars shown for some of the computational values are not error bars, but demonstrate that for these systems we had at least two data sets (see Table I); the ends of the bars mark the extreme values from the sets, and the dot marks their

TABLE I: Simulated single component fully hydrated lipid bilayer systems. The simulation file data sets marked with * also include a part of the trajectory. If simulation data from previously published work has been directly used, the original publication is cited for simulation details. For other systems the simulation details are given in the Supplementary Information. The abbreviations are 1-palmitoyl-2-oleoylphosphatidylcholine (POPC), dipalmitoylphosphatidylcholine (DPPC), 1,2-dimyristoyl-sn-glycero-3-phosphocholine (DMPC), 1,2-dioleoyl-sn-glycero-3-phosphocholine (DOPC) and dilauroylphosphatidylcholine (DLPC).

Force field	lipid	^a N _l	^b N _w	^c T (K)	^d t _{sim} (ns)	^e t _{anal} (ns)	^f Files	^g Details
Berger-POPC-07 [68]	POPC	128	7290	298	270	240	[72]*	[69]
5. Berger-DPPC-98 [73]	DPPC	72	2864	323	140	100	[74]	SI
Berger-DMPC-04 [75]	DMPC	128	5097	323	130	100	[76]	[77]
CHARMM36 [31]	POPC	72	2242	303	30	20	[78]*	SI
CHARMM36 [31]	POPC	128	5120	303	150	100	[79]* 6.	SI
CHARMM36 [31]	DPPC	72	2189	323	30	25	[80]* 7.	SI
8. 9. MacRog [37]	POPC	288	12600	310	100	80	[81]*	SI
GAFFlipid [33]	POPC	126	3948	303	137	32	[82]*	SI
GAFFlipid [33]	DPPC	72	2197	323	90	50	[83]*	SI
Lipid14 [84]	POPC	72	2234	303	100	50	[85]*	SI
Poger [86]	DPPC	128	5841	323	2×100	2×50	[87, 88]* 10.	SI
Slipid [89]	DPPC	128	3840	323	150	100	[90]*	SI
Slipid [91]	POPC	128	5120	303	200	150	[92]*	SI
Kukol [93]	POPC	512	20564	298	50	30	[94]*	SI
Chiu [95] 11.	POPC	128	3552	298	56	50	[96]*	SI
Högborg08 [97] 12.	POPC	128	3840	303	100	80	?	[97]
Högborg08 [29] 13.	DMPC	98	3840	303	75	50	?	[29]
Ulmschneiders [98]	POPC	128	3328	310	100	50	[99]*	SI
Tjörnhammar14 [100]	DPPC	144	7056	323	200	100	[101]*	[100]
CHARMM36-UA [102, 103]	DLPC	128	3840	323	30	20	[104]	SI

^aThe number of lipid molecules

^bThe number of water molecules

^cSimulation temperature

^dThe total simulation time

^eTime frames used in the analysis

^fReference link for the downloadable simulation files

^gReference for the full simulation details

calculated absolute values of the order parameters of a well-performing force field to fall.

In addition to the numerical values, an important feature of the glycerol backbone is the forking (see section II B) of the order parameters in g_1 and g_3 segments, in contrast to the choline segments α and β . The forking in glycerol backbone g_3 segment is small (≈ 0.02) and some experiments only report the larger value or the average value [35, 49]. In contrast, forking is significant for the glycerol backbone g_1 segment, whose lower order parameter is close to zero and the larger one has an absolute value of approximately 0.13–0.15. Forking was studied in detail by Gally et al. [26], who used *E. Coli* to stereospecifically deuterate the different hydrogens attached to the g_1 or g_3 groups in PE lipids, and measured the order parameters from the lipid extract. This experiment gave the lower order parameter when deuterium was in the S position of g_1 or R position for g_3 . Since the glycerol backbone order parameters are very similar irrespective of the headgroup chemistry (PC, PE and PG)

or lipid environment [26], it is reasonable to assume that the stereospecificity measured for the PE lipids holds also for the PC lipids.

By combining the absolute value measurements [35], the sign measurements [16, 63, 64] and the stereospecific measurements [26] the most detailed experimentally available order parameter information for the glycerol backbone and choline segments of POPC bilayer is collected and the data are shown in Fig. 3.

B. Full hydration: Comparison between simulation models and experiments

The order parameters of the glycerol backbone and head-group calculated from different force fields for various lipids have been previously compared to experiments [28–37]. The general conclusion from these studies seems to be that the CHARMM based [29, 31], GAFFlipid [33] and MacRog [37]

TABLE II: Simulated single component lipid bilayers with varying hydration levels. The simulation file data sets marked with * include also part of the trajectory.

Force field	lipid	^a n (w/l)	^b N _l	^c N _w	^d T (K)	^e t _{sim} (ns)	^f t _{anal} (ns)	^g Files	^h Details
Berger-POPC-07 [68]	POPC	57	128	7290	298	270	240	[72]*	SI
	POPC	7	128	896	298	60	50	[105]*	SI
CHARMM36[31]	POPC	31	72	2242	303	30	20	[78]*	SI
	POPC	15	72	1080	303	59	40	[106]*	SI
	POPC	7	72	504	303	60	20	[107]*	SI
	POPC	50	288	14400	310	90	40	[108]*	SI
MacRog[37]	POPC	25	288	7200	310	100	50	[108]*	SI
	POPC	20	288	5760	310	100	50	[108]*	SI
	POPC	15	288	4320	310	100	50	[108]*	SI
	POPC	10	288	2880	310	100	50	[108]*	SI
	POPC	5	288	1440	310	100	50	[108]*	SI
GAFFlipid[33]	POPC	31	126	3948	303	137	32	[82]*	SI
	POPC	7	126	896	303	130	40	[109]*	SI

^aWater/lipid molar ratio

^bThe number of lipid molecules

^cThe number of water molecules

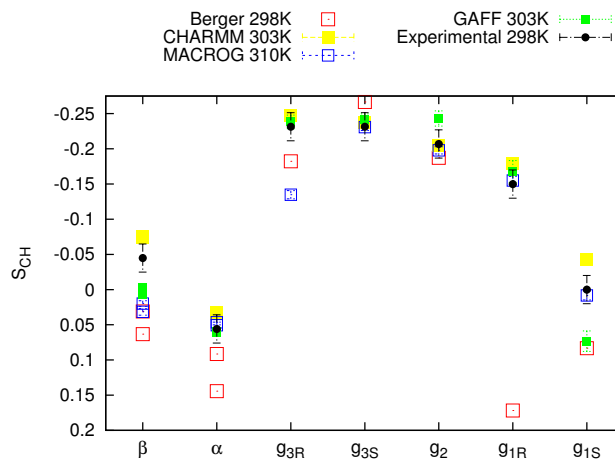
^dSimulation temperature

^eThe total simulation time

^fTime frames used in the analysis

^gReference link for the downloadable simulation files

^hReference for the full simulation details



19. Markus Miettinen suggested that we should consider “spreading” the values similarly to the figure 2

FIG. 3: Order parameters from simulations with Berger-POPC-07, CHARMM36, GAFFlipid and MacRog force fields together with experimental values for POPC glycerol and choline groups. The magnitudes for experimental order parameters are taken from Ferreira et al. [35], the signs are based on the measurements by Hong et al. [16, 63] and Gross et al. [64], and the R/S labeling is based in the measurements by Gally et al. [26].

force fields perform better for the glycerol backbone and head-group structures than the GROMOS based models [30, 32, 34, 35]. However, none of the studies exploits the full potential of the available experimental data discussed in previous section, i.e. the quantitative accuracy, known signs and stereospecific labeling of the experimental order parameters.

To get a general idea of the quality of the glycerol backbone

and choline headgroup structures in different models, we calculated the order parameters for these parts from thirteen different lipid models (Table I) and plotted the results together with experimental values in Fig. 2. Two criteria were used to judge the quality of the model: there must not be significant **forking** in the α and β carbons, there must be only moderate forking in the g_3 carbon and there must be significant

TABLE III: Simulated lipid bilayers containing cholesterol. The simulation file data sets marked with * include also part of the trajectory.

Force field	lipid	^a N _l	^b N _{chol}	^c N _w	^d T (K)	^e t _{sim} (ns)	^f t _{anal} (ns)	
Berger-POPC-07 [68]/Höltje-CHOL-13 [35, 110]	POPC	128	0	7290	298	270	240	[72]* [69]
	POPC	120	8	7290	298	100	80	[111]* [35]
	POPC	110	18	8481	298	100	80	[112]* [35]
	POPC	84	44	6794	298	100	80	[113]* [35]
	POPC	64	64	10314	298	100	80	[114]* [35]
	POPC	50	78	5782	298	100	80	[115]* [35]
CHARMM36[31, 116] 14. 15. 16.	POPC	128	0	5120	303	150	100	[79]* SI
	POPC	100	24	4960	303	200	100	[117]* SI
	POPC	80	80	4496	303	200	100	[118]* SI
MacRog[37]	POPC	128	0	6400	310	400	200	[119]* SI
	POPC	114	14	6400	310	400	200	[119]* SI
	POPC	72	56	6400	310	400	200	[119]* SI
	POPC	64	64	6400	310	400	200	[119]* SI
	POPC	56	72	6400	310	400	200	[119]* SI

^aThe number of lipid molecules
^bnumber of cholesterol molecules
^cThe number of water molecules
^dSimulation temperature
^eThe total simulation time
^fTime frames used in the analysis

forking in the g₁ carbon, the **magnitude** should be preferably inside to the subjective sweet spots determined from experiments (blue shaded regions in Fig. 2). The results for each force field in respect to the above criteria are summarized in Figure 4. **20.DONE.**

None of the studied force fields fulfils these criteria completely, however CHARMM36 is pretty close. This is not surprising since the dihedral potentials in this model are tuned to reproduce these parameters better against experiments [31]. The next models in the list are CHARMM36-UA [102, 103] and Högberg08 [29] which is also not surprising since these models are using CHARMM bonded potentials for glycerol backbone and choline. The fourth and the fifth models in the list, MacRog [37] and GAFFlipid [33], have independently determined dihedral potentials. All the models based on Gromos potentials and Slipid perform less well. In the present work we subject the CHARMM36, MacRog, GAFFlipid and Berger-POPC-07 to a more careful comparison including the stereospecific labeling (Fig. 3), and atomistic level structure and responses to the dehydration and cholesterol content in the following sections. These models are selected for more detailed studies since they are the best representatives of different dihedral potential parametrization techniques (CHARMM36, MacRog, GAFFlipid), and the Berger based models are the most used lipid model in the literature.

C. Full hydration: Atomistic resolution structures in different models

The results in the previous section revealed significant differences of the glycerol backbone and choline headgroup order parameters between different molecular dynamics simulation models. However, it is not straightforward to conclude which kind of structural differences (if any) between the models the results indicate, because the mapping from the order parameters to the structure is not unique. In this section we demonstrate that 1) the differences in order parameters indicate significantly different structural sampling strongly correlating with the dihedral angles of the related bonds, and that 2) the comparison between experimental and simulated order parameters can be used to exclude nonrealistic structural sampling in molecular dynamics simulations. The demonstration is done for the dihedral angles defined by the g₃-g₂-g₁-O(*sn*-1) segments in the glycerol backbone and the N-β-α-O segments in the headgroup. These dihedrals were chosen for demonstration, because significant differences between the models are observed around these segments in Fig. 3. We note that performing a similar comparison through all the dihedrals in all the 13 models would probably give highly useful information on how to improve the accuracy of the models yet this is beyond the scope of the current report.

The dihedral angle distributions for the g₃-g₂-g₁-O(*sn*-1) dihedral calculated from different models are shown in Fig. 5. The distribution is qualitatively different for the Berger-POPC-07 model, showing a maximum in the gauche⁺-conformation (60°) compared to all the other models show-

	β	α	g_3	g_2	g_1	Σ
CHARMM 36	M		M		M	4
CHARMM 36-UA	M	F	M F	M	M F	7
Högberg08		M F	M F			9
MacRog	M F	F	M F			11
GAFFlipid	M F		F	M	M F	11
Lipid14	M	M	M F	M	M F	16
Ulm-schneiders	M	F	M	M	M F	17
Tjörnhammar14	M	M	M F	M	M F	19
Slipid	F	F	M	M	M F	19
Poger	M F	M F	M F	M	M	23
Chiu	M F	M F	M F	M	M	23
Berger	M F	M F	M F		M F	26
Kukol	M F	M F	M F	M	M F	27

FIG. 4: Rough ranking of force fields based on data of Fig. 2. "M" indicates a magnitude problem, "F" a forking problem. Letter size shows the level (0-4) of severity; the Σ -column shows the sum of these, i.e., the "total severity". Color scheme: "within experimental error" (dark green), "almost within experimental error" (light green), "clear deviation from experiments" (light red), and "major deviation from experiments" (dark red).

ing a maximum in the anti-conformation (180°). The distributions in all the other models have the same general features, the main difference being that the fraction of configurations in the gauche⁻-conformation (-60°) is zero for the MacRog, detectable for the CHARMM36 and equally large to the gauche⁺ fraction in GAFFlipid. From the results we conclude that most likely the wrongly sampled dihedral angle for the g_2 - g_1 bond explains the significant discrepancy to the experimental order parameters for the g_1 segment in the

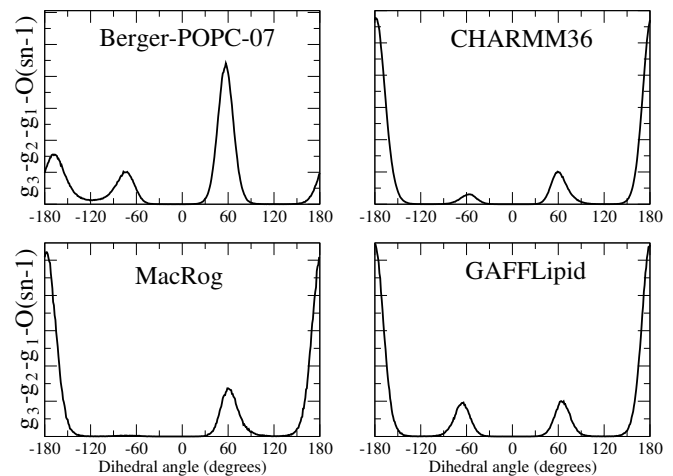


FIG. 5: Dihedral angle distributions for g_3 - g_2 - g_1 -O($sn-1$) dihedral from different models (POPC bilayer in full hydration).

Berger-POPC-07 model (Fig. 3). In conclusion, models preferring the anti conformation for this dihedral give more realistic order parameters and this is in agreement with previous crystal structure and ^1H NMR studies [19–21, 23–25].

The dihedral angle distribution for the N- β - α -O dihedral calculated from the same four models is shown in Fig. 6. Also for this dihedral there are significant differences in the gauche-anti fractions. The gauche conformations are dominant in the CHARMM36, in MacRog there are only anti conformations present, and in the Berger-POPC-07 and GAFFlipid gauche and anti conformations have equal probabilities. On the other hand, comparison of α and β order parameters in Fig. 3 reveals that for these carbons the CHARMM36 is closest to the experimental results and it is also the only model that has the correct sign (negative) for the β order parameter. This result is again in agreement with previous crystal structure, ^1H NMR and Raman spectroscopy studies [19–22] which suggest that this dihedral is in the gauche conformation in the absence of ions.

The used examples show that the glycerol backbone and headgroup order parameters reflect the atomistic resolution structure and that the comparison with experiments allows the assessment of the quality of the suggested structure. We were able to pinpoint specific problems in the structures in different models and suggest potential improvement strategies. If the improved atomistic molecular dynamics simulation model reproduced the order parameters and other experimental observables (like chemical shift anisotropy) with experimental accuracy, it would give an interpretation for the atomistic resolution structure of the glycerol backbone and choline [10–13, 15, 16, 18]. The research along these lines is left, however, for future studies.

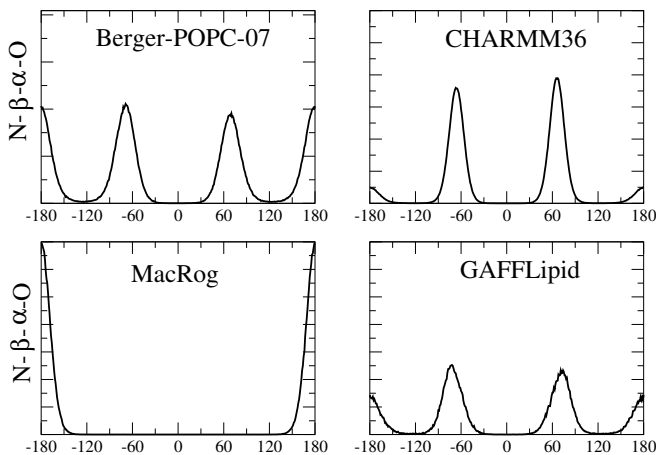


FIG. 6: Dihedral angle distributions for N- β - α -O dihedral from different models (POPC bilayer in full hydration).

D. Response to dehydration and cholesterol content

In addition to pure phosphatidylcholine bilayers at full hydration, the choline headgroup order parameters have been measured under various different conditions [30, 32, 35, 44–50, 53, 54]. Also the order parameters for the glycerol backbone have been measured with ^{13}C NMR in dehydrated conditions [46], and as a function of anesthetics [30] and glycolipids [32] for DMPC, and as a function of cholesterol concentration for POPC [35]. Due to the high resolution in the NMR (especially ^2H NMR) experiments, even very small order parameter changes resulting from the varying conditions can be measured (see <http://nmrlipids.blogspot.fi/2014/02/accuracy-of-order-parameter-measurements.html> for more discussion.) **21. This is temporarily linked to the blog. Currently I think that the posts (with comments included) to which we want to cite are put into the Figshare in pdf format as they are in when the publication is submitted. Figshare allows commenting to continue the discussion.** However, as already discussed above, it is not simple to deduce the structural changes from order parameter changes [15, 18]. Consequently, comparison of the order parameters between simulations and experiments in different conditions can be used to assess the quality of the force field in different situations, and, if the quality is good, to potentially interpret the structural changes in experiments. Here we exemplify such comparison for a lipid bilayer under low hydration levels and when varying amounts of cholesterol is included in the bilayer. The interaction between ions and a phosphatidylcholine bilayer will be discussed in a separate study [59].

1. Phospholipid bilayer with low hydration level

The experimental order parameters available in the literature [44–46] for the glycerol backbone and choline as a function of hydration level are shown in Fig. 7. The independently reported values for choline segments are in good agree-

ment with each other (despite slight differences in temperature and acyl chain composition), showing increase for both segments with decreasing hydration level. It should be noted that only absolute values were measured in the original experiments [44–46], but we have included the signs measured in separate studies [16, 63, 64]. Consequently, the β order parameter with negative sign actually increases with dehydration since the absolute value decreases [44–46]. Slight decrease for the glycerol backbone g_3 - and g_2 - order parameters were observed with dehydration, while g_1 remained practically unchanged [46].

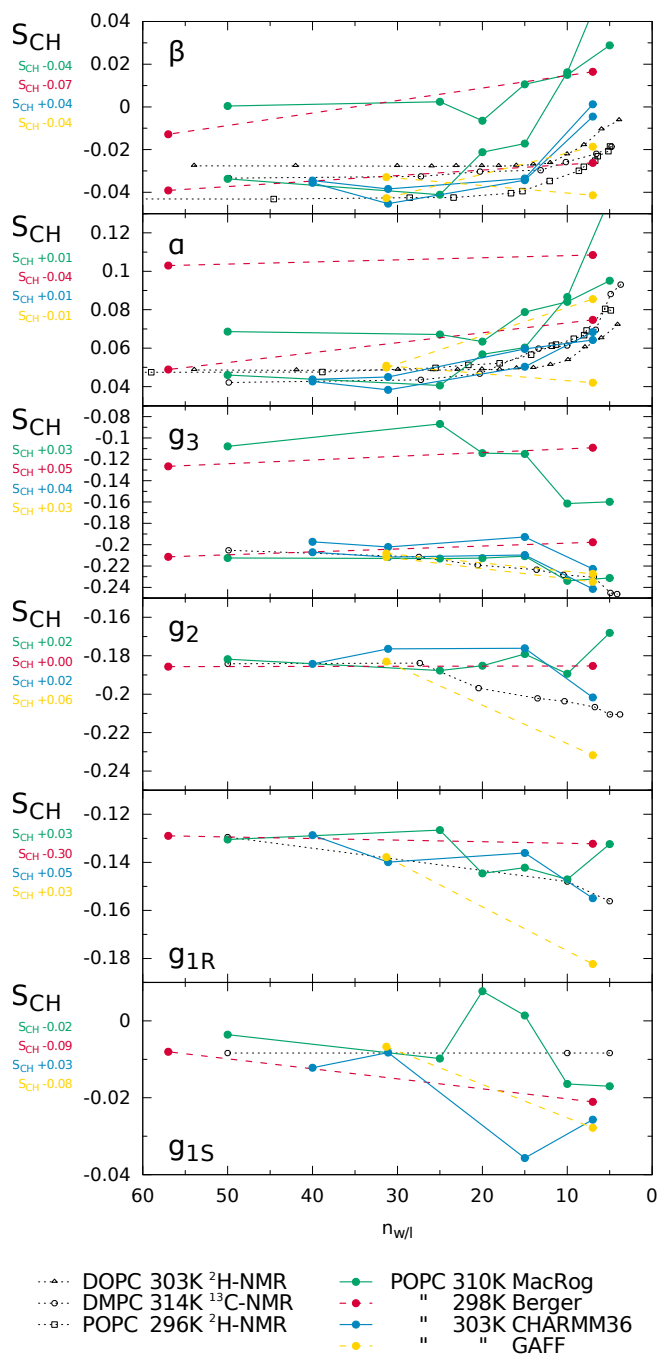
Lipid bilayer dehydration has been studied also with molecular dynamics simulations [121–126], typically motivated by the discussion about the origin of the “hydration repulsion” [127–129]. However, the used simulation models are not typically compared to the experimental choline and glycerol backbone order parameters (except by Mashl et al. [121]). In Fig. 7 the glycerol backbone and choline order parameters are shown as a function of hydration level for the CHARMM36, MacRog and GAFFlipid models (having the most realistic atomistic resolution structures) together with the Berger based model (which is the most used lipid model). The choline order parameters increase with dehydration in all models, in qualitative agreement with experiments. The measured decrease in both g_3 and g_2 order parameters with dehydration is reproduced only in CHARMM36.

The qualitative agreement with experiments in all simulation models for the α and β order parameters as a function of hydration indicates that the structural response of the choline headgroup to dehydration is somewhat realistic despite the unrealistic structures at full hydration. The most likely explanation is that the choline group orients more parallel to the membrane plane with dehydration due to restricted interlamellar space. Indeed, the P–N (phosphate phosphorus to choline nitrogen) vector angle with membrane normal shows an increase for all models as a function of dehydration in Fig. 8. However, the qualitative agreement in the lipid response to dehydration does not guarantee the correct free energy landscape if the simulation model has an incorrect structure. The influence of this issue to dehydration energetics studied with simulations [124, 126] is left for future studies.

The response of the glycerol backbone to dehydration seems to be more subtle than that of the choline headgroup as CHARMM36 is the only model that reproduces the decrease in g_2 and g_3 segments.

2. Cholesterol-containing phospholipid bilayer

Phospholipid–cholesterol interactions have been widely studied with theoretical [130–133] and experimental methods [8, 35, 47, 134], since cholesterol is abundant in biological membranes. It has been suggested to be an important player, for example, in domain formation [135, 136]. It is widely agreed that cholesterol orders lipid acyl tails thus decreasing the area per molecule (condensing effect), however, the influence of cholesterol on the lipid headgroup and glycerol backbone are still debated [130, 135, 136]. For exam-



22. Figure has been modified to elucidate changes. The text should be updated accordingly. The same applies to the cholesterol figure.

FIG. 7: The effect of dehydration on glycerol and choline order parameters in experiments. The magnitudes of order parameters are measured for DMPC (^{13}C NMR) at 314 K [46], for POPC (^2H NMR) at 296 K [44] and for DOPC (^2H NMR) at 303 K [45]. The signs are based on the measurements by Hong et al. [16, 63] and Gross et al. [64].

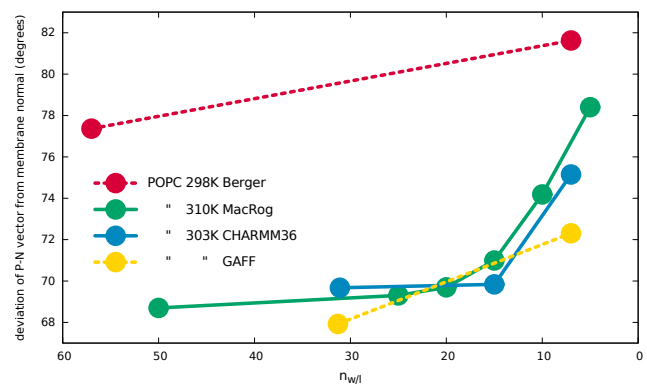


FIG. 8: The angle between membrane normal and P-N vector as function of hydration level calculated from different simulations.

ple, it has been suggested that the surrounding lipids shield cholesterol from interactions with water by reorienting their headgroups (“umbrella model”) [130] or that cholesterol acts as a spacer for the headgroups thus increasing their entropy and dynamics (“superlattice model”) [136]. **23. It has been suggested that we should use direct quotations from original papers here for clarity.** Both of these suggestions have been supported by molecular dynamics simulations [131, 133], and other simulations suggest specific interactions between the glycerol backbone and cholesterol [132], however the glycerol backbone and choline headgroup behaviour as a function of cholesterol content is not compared to experiments in these studies.

The choline headgroup and glycerol backbone order parameters for POPC measured by ^{13}C NMR [35] and DPPC choline order parameters measured by ^2H NMR [47] are shown in Fig. 9 as a function of cholesterol content. The agreement between different experimental results is again very good, showing only very modest changes in the choline order parameters as a function of cholesterol content. It should be noted, however, that very small changes are measurable with high resolution ^2H NMR experiments and cholesterol causes a measurable increase in the β order parameter and a forking in the α order parameter [47]. These effects are, however, so small that they are barely visible in the scale used in Fig. 9. Further, the effects of cholesterol on the glycerol backbone order parameters for POPC from ^{13}C NMR experiment [35] are in good agreement with the results for the phosphatidylethanolamine (PE) measured by ^2H NMR [137]. These results further support the idea that the glycerol backbone structural behaviour is independent of the headgroup composition [26] and that the headgroup structure is independent of the acyl chain region content unless charges are present [27].

In addition to the experimental data, the previously published simulation results from the Berger-POPC-07/Höltje-CHOL-13 model [35], and our results from CHARMM36 and MacRog force fields are shown in Fig. 9. As already pointed out previously, the Berger-POPC-07/Höltje-CHOL-13 model seriously overestimates the effect of cholesterol

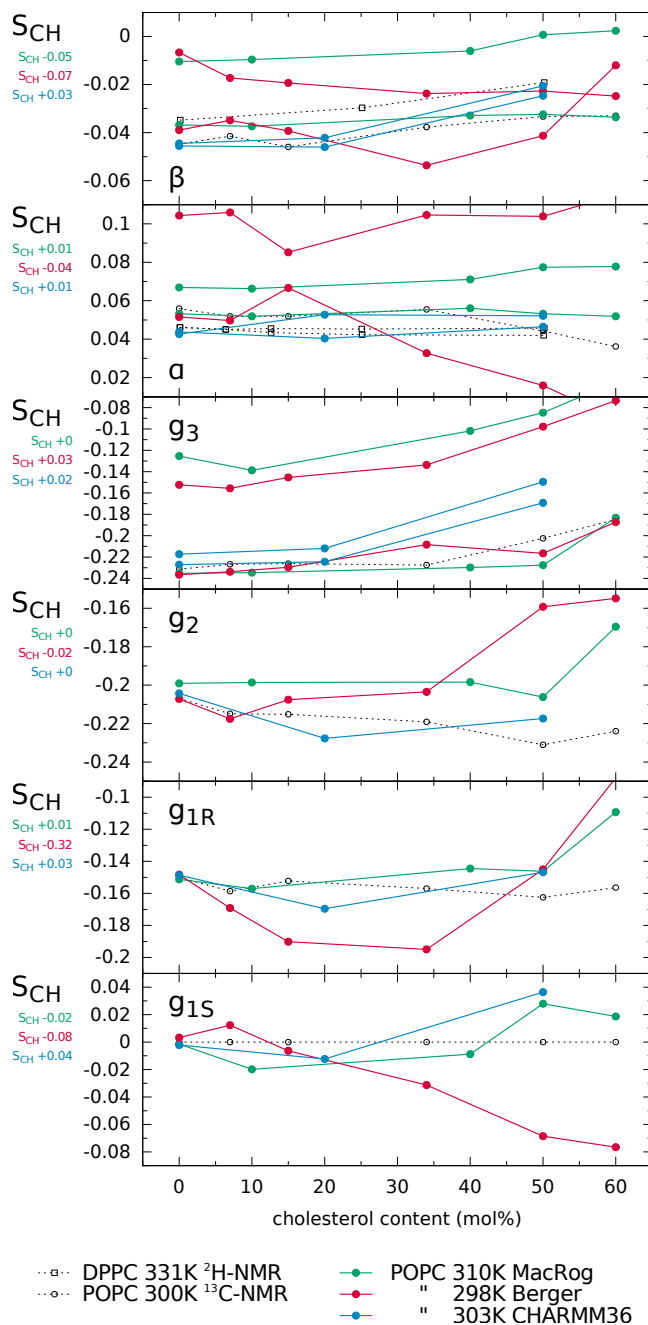


FIG. 9: The effect of cholesterol content on the glycerol backbone and choline order parameters in experiments [35, 47] and simulations with the Berger-POPC-07/Höltje-CHOL-13, CHARMM36 and MacRog force fields. The signs in the experimental values are based on the measurements by Hong et al. [16, 63] and Gross et al. [64]. Most order parameters from Berger-POPC-07/Höltje-CHOL-13 model for g_1 are beyond the y-axis scale.

on the phospholipid glycerol backbone and choline segments [35]. In contrast, the responses of both CHARMM36 and MacRog are in better agreement with experiments. However, CHARMM36 seems to better reproduce the experimentally observed modest changes in the glycerol backbone segments g_2 and g_3 with high concentrations of cholesterol. Thus we have calculated the glycerol backbone dihedral angle distributions as a function of cholesterol in CHARMM36 (shown in Supplementary material) to resolve the cholesterol induced structural changes. The only detectable change due to the addition of cholesterol is the small decrease of gauche- and increase of gauche+ probability of g_3 - g_2 - g_1 -O(*sn*-1) dihedral.

It should be noted that the CHARMM36 force field parameters (dihedral potentials) for the glycerol backbone have been tuned to reproduce the correct order parameters at fully hydrated conditions [31]. This procedure contains a risk of overfitting, which would manifest itself as wrong responses to changing conditions. According to our results, this tuning does not seem to lead to overfitting problems in the case of dehydration or lipid-cholesterol mixtures.

IV. CONCLUSIONS

The atomistic resolution structures sampled by the glycerol backbone and choline headgroup in phosphatidylcholine bilayers are not known despite of vast amount of accurate experimental data. Atomistic resolution molecular dynamics simulation model which reproduced the experimental data would automatically resolve the structures giving an interpretation of experimental results. In this work we have collected and reviewed the experimental C–H bond vector order parameters available in the literature. These experimental parameters are then compared to different atomistic resolution simulation models for a fully hydrated bilayer, bilayers dehydrated to different extents, and lipid bilayers containing various amounts of cholesterol. Our results have led to the following conclusions:

- The C–H bond order parameters measured with different NMR techniques are in good agreement with each others. By combining the experimental results from various sources we concluded that the order parameters for each C–H bond are known with a quantitative accuracy of ± 0.02 .

- None of the tested models (13 different models) produces the order parameters with the experimental accuracy for a fully hydrated phosphatidylcholine lipid bilayer. However, the CHARMM36, GAFFlipid and MacRog models are relatively close. The structures of these models together with the most used lipid model (Berger based) were subjected to more careful studies. The results revealed that the current models are not accurate enough to resolve the atomistic resolution structures sampled by glycerol backbone and choline headgroup. However, the correlation between dihedral angle distributions and order parameter differences was found, suggesting that careful adjustment of dihedral potentials would potentially lead to the model with a correct structure.

- Independent of the accuracy for a fully hydrated lipid bilayer, all the models reproduced the choline response to dehydration. This can be explained by the change in the P–N vector tilting more parallel to the membrane, which leads to the increase of order parameters despite of the initial configuration. It should be noted, however, that the correct qualitative response does not necessarily indicate correct energetics.

- The response of glycerol backbone and choline headgroup to the cholesterol content is described more realistically in CHARMM36 and MacRog models than in the Berger based model.

- **Besides the main conclusions, we have created the most extensive publicly available collection of molecular dynamics simulation trajectories of lipid bilayers into the Zenodo community <https://zenodo.org/collection/user-nmrlipids>. This collection opens up numerous possibilities for different analyses with much less effort than previously required.**

In general, we conclude that to fully utilize the potential of atomistic resolution classical molecular dynamics simulations in the structural interpretation of the high resolution NMR data [138] for lipid bilayers one has to improve the phosphatidylcholine glycerol backbone and choline headgroup parameters.

This work has been done as an open collaboration by using the `nmrlipids.blogspot.fi` as a communication platform. All the scientific contributions have been done through the blog and are publicly available.

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SUPPLEMENTARY INFORMATION

1. Simulation details

a. Berger based models

For the Berger based models we use here the following naming convention: Berger - {*molecule name*} - {*year when model published first time*} {*citation*}. The reason is that there are several different molecular topologies which are using the non-bonded parameters originally developed by Berger et al. [43]. Thus the common factor in the Berger based models are the non-bonded parameters, while the molecule specific parameters might somewhat vary. However, the majority of the molecular level topologies are relying (especially for the glycerol backbone and headgroup) on the parameters originally introduced by Marrink et al. [73]. This is the case for all the Berger based simulations discussed in this work.

POPC simulations at full hydration at 298 K and simulations studying the effect of cholesterol are the same as in previous publications [35, 69]. In these simulation the POPC parameters introduced by Ollila et al [68] are used, which are using the non-bonded parameters of Berger [43] and a molecular topology from Tieleman et al. [139] with improved double bond dihedrals by Bachar et al. [140]. Thus they are called Berger-POPC-07 [68]. The cholesterol model is based on the parameters by Höltje et al. [110] with the exception that the atom types were changed from CH₂/CH₃ to LP2/LP3 to avoid overcondensation of the bilayer as suggested in ref. [141]. Since this modification was introduced by Ferreira et al. [35], we call the used cholesterol model as Höltje-CHOL-13 [35].

For the POPC at 323 K and POPC in low hydration the same force field parameters are used. For DPPC the implementation of Berger parameters [43] by Peter Tieleman et al. are used [73]. For all of these simulations a timestep of 2 fs was used with a leapfrog integrator. Covalent bond lengths were constrained with the LINCS algorithm [142, 143]. Coordinates were written every 10 ps. PME [144, 145] with real space cut-off of 1.0 nm was used for electrostatics. Plain cut-off was used for the Lennard-Jones interactions with a 1.0 nm cut-off. The neighbor lists with cut-off of 1.0 nm were updated every 5 steps. Temperature was coupled separately for lipids and water to 298 K using the velocity-rescale method [146] with coupling constant 0.1 ps. Pressure was semi-isotropically coupled to the atmospheric pressure with the Berendsen method [147].

b. CHARMM36

DPPC 24. Markus Miettinen made the files.

Timestep of 1 fs was used with the leapfrog integrator. Covalent bonds with hydrogens were constrained with LINCS algorithm [142, 143]. Coordinates were written every 5 ps. PME [144, 145] with real space cut-off of 1.4 nm was used for electrostatics. Lennard-Jones interactions were switched to zero between 0.8 nm and 1.2 nm. The neighbour lists with

a cut-off of 1.4 nm were updated every 5 steps. Temperature was coupled separately for lipids and water to 303 K using the velocity-rescale method [146] with coupling constant of 0.2 ps. Pressure was semi-isotropically coupled to the atmospheric pressure with the Berendsen method [147].

POPC The starting structures for the pure POPC and DOPC simulations was taken from the Slipids [91] website (http://people.su.se/~jjm/Stockholm_Lipids/Downloads.html). The starting structures for mixed POPC/Cholesterol simulations were constructed with the CHARMM-GUI website [148]. They contained 100 POPC/24 cholesterol molecules and 80 POPC/80 cholesterol molecules for the simulations of 20% cholesterol and 50% cholesterol respectively. The TIP3P water model [149] was used to solvate the system. The publicly available CHARMM36 forcefield parameters (http://www.gromacs.org/@api/deki/files/184/=charmm36.ff_4.5.4_ref.tgz) by Piggot et al. [6] were used. Cholesterol parameters came from Lim et al. [116] and were converted into GROMACS format with the PyTopol tool (<https://github.com/resal81/PyTopol>). Single point energy calculation was done to assess the conversion. Simulations were performed for 200 ns and the last 100 ns was used for the calculations. Timestep of 2 fs was used with leapfrog integrator. All bond lengths were constrained with LINCS [142, 143]. Temperature was maintained at 303 K with the velocity-rescale method [146] and a time constant of 0.2 ps. Pressure was maintained semiisotropically at 1 bar using the Parrinello–Rahman algorithm [150] with a time constant of 1.0 ps. The neighbour list with a cut-off of 1.2 nm was updated every 10 steps. Lennard-Jones interactions were switched to zero between 0.8 nm and 1.2 nm. PME [144, 145] with real space cut-off of 1.2 nm was used for electrostatics.

c. MacRog

The lipid force field parameters were obtained from the developers and they correspond to the published DPPC parameters [37] with the inclusion of the double bond parameters. This inclusion of unsaturated lipid tails will be published in the near future. **25. You have recent paper where you use POPC/chol simulations with this model: Kulig et al. BBA 1848 (2015) 422-432 <http://dx.doi.org/10.1016/j.bbamem.2014.10.032>. would this be a correct refence here?** A bilayer with 288 POPC lipids was hydrated with 12600 TIP3P water [149] molecules (~ 44 /lipid) and simulated for 100 ns with a time step of 2 fs. Data was saved every 10 ps and the first 20 ns of the trajectory was discarded from the analysis. All bond lengths were constrained with LINCS [142, 143]. The temperatures of the lipids and the solvent were separately coupled to the Nosé–Hoover thermostat [151, 152] with a target temperature of 310 K and a time constant of 0.4 ps. Semi-isotropical pressure coupling to 1 bar was obtained with the Parrinello–Rahman barostat [150] with a time constant of 1 ps. PME [144, 145] was employed to calculate the long-range electrostatic interactions. Lennard-Jones interactions were cut off at 1 nm and the dispersion correction was applied to both energy and pressure. A neighbour

list with a radius of 1 nm was updated every step.

Identical parameters were employed for both full hydration and for the dehydration simulations. The dehydration simulations were also run for 100 ns with data saved every 10 ps.

The initial structures for the simulations with 10, 40, 50 and 60 mol% of cholesterol were obtained by replacing 14, 56, 64 or 72 POPC molecules with cholesterol molecules in the initial structure containing 128 POPC molecules. These systems were simulated for 400 ns and the first 200 ns was discarded from analysis. Data was saved every 100 ps.

d. GAFFLipid

The initial structure in Lipidbook [153] had different glycerol backbone isomers in different leaflets. To generate the initial structure we took the structure delivered by Slipids developers [91]. Also this structure had one lipid with different glycerol backbone isomer. This lipid and one lipid from opposite leaflet were removed after the system was equilibrated.

The force field parameters were generated using files obtained from the Lipidbook website (<http://lipidbook.bioch.ox.ac.uk/package/show/id/150.html>) [153]. The conversion to GROMACS compatible formats was performed using the acpype tool [154]. The accuracy of the conversion was checked by calculating the total energy of a single POPC lipid molecule using the sander program which is part of the AmberTools14 package [155] and version 4.6.5 of GROMACS. A difference of 0.002 kcal/mol was obtained between the two programs.

Timestep of 2 fs was used in Langevin dynamics with zero friction term and collision frequency of 1.0 ps^{-1} . Covalent bonds with hydrogens were constrained with the LINCS algorithm [142, 143]. Coordinates were written every 10 ps. PME [144, 145] with a real space cut-off at 1.0 nm was used for electrostatics. Plain cut-off with 1 nm was used for Lennard-Jones interactions. The neighbour lists with a cut-off of 1.0 nm were updated every 5 steps. Pressure was semi-isotropically coupled to a pressure of 1 bar with the Berendsen method [147].

It should be noted that the area per molecule with these settings for the GAFFlipid model was 61.6 \AA^2 , while the original publication reported 63.9 \AA^2 [33]. However, the same parameters and Amber to Gromacs conversion reproduced the area per molecule from original publication for the lipid14 model (see next section).

e. Lipid14

The initial structure was taken directly from the Lipidbook [153]. The Amber compatible force field parameters were generated using the tleap program which is integrated in the AmberTools14 package [155]. A workflow similar to the one used previously for the conversion and validation of the GAFFLipid parameters was followed here. As before, a negligible energy difference of 0.003 kcal/mol was obtained between the two programs.

Timestep of 2 fs was used in Langevin dynamics with zero friction term and collision frequency of 1.0 ps^{-1} . Covalent bonds with hydrogens were constrained with LINCS algorithm [142, 143]. Coordinates were written every 10 ps. PME [144, 145] with real space cut-off of 1.0 nm was used for electrostatics. Plain cut-off with 1 nm was used for Lennard-Jones interactions. Dispersion correction was applied for both energy and pressure. The neighbor lists with a cut-off of 1.0 nm were updated every 5 steps. Pressure was semi-isotropically coupled to a pressure of 1 bar with the Berendsen method [147].

The area per molecule with these settings was 65.4 \AA^2 which is in agreement with the value reported in the original publication $65.6 \pm 0.5 \text{ \AA}^2$ [84].

f. Poger et al.

The Poger lipids are derived from GROMOS G53A6 [86] and were initially coined 53A6-L (L for lipids). They are now part of GROMOS G54A7 [34] and parametrized to work with the SPC water model [156]. The initial hydrated bilayer structure of 128 DPPC and 5841 water molecules as well as force field parameters were downloaded from David Poger's web site (<http://compbio.chemistry.uq.edu.au/~david/>) on April 2012. We noticed that the same files downloaded in October 2013 appear to lack two dihedral angles in the choline headgroup (only one dihedral of type gd.29 allowing the rotation of the 3 choline methyls) compared to the April 2012 version (3 dihedrals of type gd.29 for the 3 choline methyls). This should not affect the bilayer structure and only change the kinetics of the choline methyls rotation. However the October 2013 version has not been tested in this study.

MD Simulations (two repetitions with independent initial velocities) were run for 100 ns using a 2 fs time step and the analysis was performed on the last 50 ns. Coordinates were saved every 50 ps for analysis. All bond lengths were constrained with the LINCS algorithm [142, 143]. Temperature was kept at 323 K employing the velocity-rescale [146] thermostat with a time constant of 0.1 ps (DPPC and water coupled separately). Pressure was maintained semi-isotropically at 1 bar using the Parrinello–Rahman barostat [150] using a 4 ps time constant and a compressibility of $4.5 \times 10^{-5} \text{ bar}^{-1}$. For non-bonded interactions, two conditions were tested: i) A 0.8–1.4 nm twin-range cut-off with the neighbor list updated every 5 steps for both electrostatics and Lennard-Jones (LJ) interactions (simulation files available at [157, 158]). For the former the generalized reaction field (RF) with a dielectric permittivity of 62 was used beyond the 1.4 nm cut-off [159]. This is the original setup that Poger et al. [86] used. ii) PME [144, 145] electrostatics with a real space cut-off of 1.0 nm, a Fourier spacing of 0.12 nm and an interpolation order of 4, LJ interactions computed with a 1.0–1.4 nm twin-range cut-off, neighbor list updated every 5 steps (simulation files available at [87, 88]). Note that Poger and Mark tested the effect of PME vs RF in ref. [34], but used a 1.0 nm cut-off with PME and 1.4 nm with RF for LJ interactions. Since 0.8–

1.4 nm twin-range cut-off for LJ interactions is used in the parametrization of GROMOS force field, we decided to use that also in the simulations with PME.

Since Poger lipids come from the GROMOS force field, it is important to note that GROMOS uses the RF scheme for computing electrostatics (this is the method used for the force field parameterization). Using setup i) based on RF, we were able to reproduce the results (i.e. area per lipid value of 0.63 nm^2) from the original work only with GROMACS versions 4.0.X and earlier (the original authors [86] used GROMACS version 3.3.3). When switching to versions 4.5.X and above, the area per lipid dropped to below 0.58 nm^2 . The GROMACS developers were contacted and a redmine issue opened (<http://redmine.gromacs.org/issues/1400>). The difference comes from the new Trotter decomposition introduced in versions 4.5.X. A fix has been introduced in version 4.6.6 that allows a recovery of an area per lipid value of 0.615 nm^2 . The results in terms of area per lipid using the different GROMACS versions are available at [158]²⁶. Thus we decided to use only the PME setup ii) for computing the order parameter since it gives stable results regardless of the GROMACS version. We obtained an area per lipid of 0.615 nm^2 , below 0.648 nm^2 found by the original authors with their PME setup (see [34]). We explained that by the fact that we used a 1.4 nm for the LJ cut-off whereas a value of 1.0 nm was used in the original publication.

g. Slipids

Initial coordinates for a hydrated DPPC (at 323 K) and POPC (at 310 K) bilayers (30 and 40 waters/lipid, respectively) were taken directly from the Slipids home page http://people.su.se/~jjm/Stockholm_Lipids/Downloads.html. The Slipids force field [89, 91] was used for the all atom descriptions of DPPC and POPC, and water was described with the TIP3P water model [149]. Simulations were performed within the NPT ensemble using the GROMACS 4.6.X simulation package [70]. The Nosé–Hoover thermostat [151, 152] was used with reference temperatures of 323 K (DPPC) and 310 K (POPC) and a relaxation time constant of 0.5 ps. Water and lipids were coupled separately to the heat bath. Pressure was kept constant at 1.013 bar using a semi-isotropic Parrinello–Rahman barostat [150] with a time constant of 10.0 ps. Equations of motion were integrated with the leapfrog algorithm using a timestep of 2 fs. Long range electrostatic interactions were calculated using the PME method [144, 145], with a fourth order smoothing spline. A real space cut-off of 1.0 nm was employed with grid spacing of 0.12 nm in the reciprocal space. Lennard-Jones potentials were cut off at 1.4 nm, with a dispersion correction applied to both energy and pressure. All covalent bonds in lipids were constrained using the LINCS algorithm [142], whereas water molecules were constrained using SETTLE [160]. Twin-range cutoffs, 1.0 nm and 1.6 nm, were used for the neighbor lists with the longrange neighbor list updated every 10 steps. This simulation protocol corresponds to the protocol

used in Ref [161].

h. Kukol

A bilayer patch with 512 POPC lipids was constructed and hydrated with ~ 40 SPC water molecules per lipid. The force field parameters were obtained from Lipidbook [153]. This bilayer was simulated with a 2 fs time step for a total of 50 ns and coordinates were saved every 100 ps. All bonds were constrained with LINCS [142, 143]. PME [144, 145] was employed for the long-range electrostatics. Lennard-Jones interactions were cut off at 1.4 nm. A neighbour list with a radius of 0.8 nm was updated every 5 steps. The constant temperature of 298 K was maintained with the Berendsen thermostat [147] with a time constant of 0.1 ps. The Berendsen barostat [147] was employed for semi-isotropic pressure coupling at 1 bar.

i. Chiu et al.

The force field parameters and the initial configuration were available through the Lipidbook [153]. Timestep of 2 fs was used with leapfrog integrator. Covalent bond lengths were constrained with LINCS algorithm [142, 143]. Coordinates were written every 10 ps. PME [144, 145] with real space cut-off of 1.0 nm was used for electrostatics. Twin range cut-off was used for the Lennard-Jones interactions with short and long cut-offs of 1.0 nm and 1.6 nm, respectively. The neighbour lists with a cut-off of 1.0 nm were updated every 5 steps. Temperature was coupled separately for lipids and water to 298 K with the velocity-rescale method [146] with a coupling constant 0.2 ps. Pressure was semi-isotropically coupled to the atmospheric pressure with the Parrinello–Rahman method [150].

j. Ulmschneider

The initial structure containing 128 POPC molecules with 3328 TIP3P water [149] molecules (26 per lipid) was downloaded from Lipidbook [153] together with the topologies. This bilayer was simulated for 100 ns with a time step of 2 fs and the data was saved every 10 ps. The bonds involving hydrogen atoms were constrained with LINCS [142, 143]. The temperature was kept at 298 K with the Berendsen thermostat [147]. The pressure was semi-isotropically coupled to the Berendsen barostat [147] with a time constant of 1 ps and a target pressure of 1 bar. PME [144, 145] was employed for long range electrostatics and a cut-off of 1 ns was employed

for the Lennard-Jones interactions. A neighbour list with a radius of 1 nm was updated every 10 steps.

Additionally, the simulations were repeated with the dispersion correction applied to pressure and temperature. Even though the area per lipid decrease d slightly, the headgroup order parameters were only slightly affected.

k. Tjörnhammar et al.

The gel phase DPPC bilayer structure delivered by Tjörnhammar and Edholm [100] was ran for 5 ns at 343 K in order to destroy the ordered gel configuration. This was followed by a 200 ns simulation at 323 K, i.e. in the fluid phase. The last 100 ns of this simulation was used for analysis. The same mdp file as in the Supplementary Information section of the original paper [100] was used except for the simulation temperature.

l. CHARMM36-UA

A hydrated bilayer consisting of 128 DLPC lipids and 3840 water molecules is modeled by the force field of Lee and co-workers [103]. This force field is a combination of the all-atom CHARMM36 force-field [31] and the united-atom Berger model [43]. The non-bonded interactions are calculated using an atom-based switching function with short and long cut-offs of 0.8 and 1.2 nm [103]. Long range electrostatic interactions are implemented using the particle-particle particle-mesh solver with a relative accuracy of 10^{-4} . The system is first equilibrated for 30 ns in the NP γ T ensemble (Nosé–Hoover [151, 152] style thermostat and barostat with anisotropic pressure coupling) at 323 K and 1 bar with timestep of 1 fs. The following 20 ns of dynamics are taken for calculation of configurational averages. Simulations were carried out by using the LAMMPS package [71].

2. Dihedral angle distributions as a function of cholesterol in CHARMM36

3. General issues to think of about the manuscript

27.DONE

28.Maybe we should mention that CHARMM is significantly more computationally expensive than other all-atom models. This seems evident for a person who has ran CHARMM and other all-atom models with the same computer. However, there may be some people who have never used any other model than CHARMM. In this case it may be difficult to realize. Would anyone have a good benchmark about this? I have assumed that the reason for the slowness are the extra Lennart-Jones interactionsi in water, however, I have never looked into it. Would someone have some hard facts on this?

TODO

- | | |
|---|-----------|
| | P. |
| 1. If there are no objections, we will leave the naming convention of different Berger versions as it is now. | 2 |
| 2. Should we write more about this? | 2 |

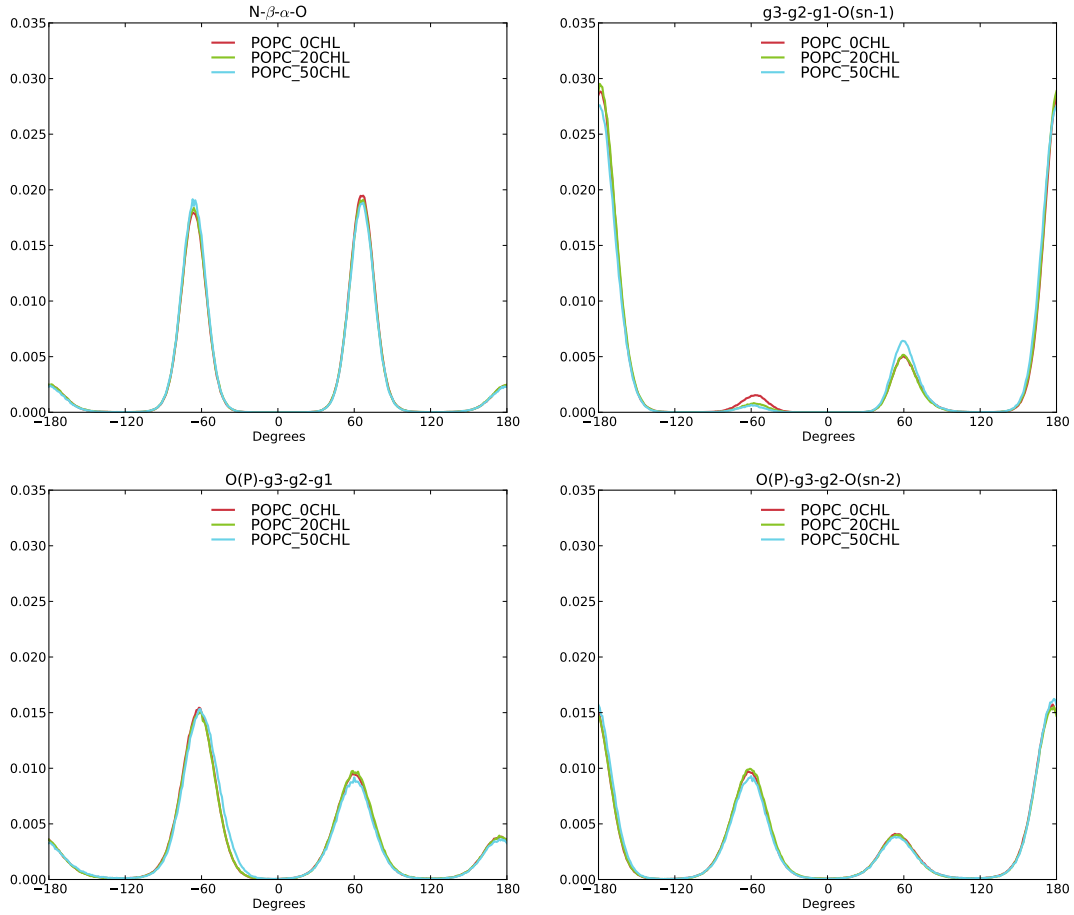


FIG. 10: The effect of cholesterol content on the glycerol backbone and choline dihedral angles in CHARMM36 model.

3. DONE	3
4. This is temporarily linked to the blog. Currently I think that the posts (with comments included) to which we want to cite are put into the Figshare in pdf format as they are in when the publication is submitted. Figshare allows commenting to continue the discussion. COMMENT: There is also the option to put the most relevant blog posts as a supplementary material. The advantage of this would be that the posts would be peer reviewed as well.	3
17. Markus Miettinen suggested that we should consider making one more figure where only experimental data would be shown and that would be discussed in Section III A. COMMENT: I do not think that we need such figure anymore since this figure is more clear now.	4
18. I think that this figure is now very good. There is also the interactive version by Hubert Santuz now in https://plot.ly/ HubertSantuz/72/lipid-force-field-comparison/ we should figure out which is the most practical way to put that behind permalink once it finalized (Zenodo, figshare or something else?) and then put a citation in the paper.	4
5. DONE	5
6. DONE	5
7. DONE	5
8. DONE	5
9. Permanent link for DOPC CHARMM36 simulation by Hubert Santuz is not required now. However, it would be probably useful in the future.	5
10. DONE	5
11. DONE	5

12. Alexander Luybartsev, let us know if you share the files.	5
13. Alexander Luybartsev, let us know if you share the files.	5
19. Markus Miettinen suggested that we should consider “spreading” the values similarly to the figure 2	6
14. DONE	7
15. DONE	7
16. DONE	7
20. DONE	7
21. This is temporarily linked to the blog. Currently I think that the posts (with comments included) to which we want to cite are put into the Figshare in pdf format as they are in when the publication is submitted. Figshare allows commenting to continue the discussion.	9
22. Figure has been modified to elucidate changes. The text should be updated accordingly. The same applies to the cholesterol figure.	10
23. It has been suggested that we should use direct quotations from original papers here for clarity.	10
24. Markus Miettinen made the files.	15
25. You have recent paper where you use POPC/chol simulations with this model: Kulig et al. BBA 1848 (2015) 422-432 http://dx.doi.org/10.1016/j.bbamem.2014.10.032 . would this be a correct refence here?	16
26. DONE	17
27. DONE	18
28. Maybe we should mention that CHARMM is significantly more computationally expensive than other all-atom models. This seems evident for a person who has ran CHARMM and other all-atom models with the same computer. However, there may be some people who have never used any other model than CHARMM. In this case it may be difficult to re-alize. Would anyone have a good benchmark about this? I have assumed that the reason for the slowness are the extra Lennart-Jones interactions in water, however, I have never looked into it. Would someone have some hard facts on this?	18