

Digitizing Poly- l -lysine Dendrigrafts: From Experimental Data to Molecular Dynamics Simulations

Jean-Patrick Francoïa, Jean-Christophe Rossi, Gerald Monard, Laurent Vial

▶ To cite this version:

Jean-Patrick Francoïa, Jean-Christophe Rossi, Gerald Monard, Laurent Vial. Digitizing Poly-1-lysine Dendrigrafts: From Experimental Data to Molecular Dynamics Simulations. Journal of Chemical Information and Modeling, 2017, 57 (9), pp.2173 - 2180. 10.1021/acs.jcim.7b00258. hal-01814838

HAL Id: hal-01814838

https://hal.science/hal-01814838

Submitted on 13 Jun 2018

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Digitizing Poly-L-lysine Dendrigrafts: from Experimental Data to Molecular Dynamics Simulations

Jean-Patrick Francoia,† Jean-Christophe Rossi,‡ Gerald Monard,*,¶,§ and Laurent Vial*,‡,∥

†Institut des Biomolécules Max Mousseron, UMR 5247 CNRS – Université de Montpellier

– ENSCM, Place Eugéne Bataillon, 34296 Montpellier cedex 5, France

‡Institut des Biomolécules Max Mousseron, UMR 5247 CNRS – Université de Montpellier

- ENSCM, Place Eugène Bataillon, 34296 Montpellier cedex 5, France

¶ Université de Lorraine, UMR 7565 SRSMC, Boulevard des Aiguillettes B.P. 70239, F-54506 Vandoeuvre-les-Nancu. France

§CNRS, UMR 7565 SRSMC, Boulevard des Aiguillettes B.P. 70239, F-54506 Vandoeuvre-les-Nancy, France

|| Institut de Chimie et Biochimie Moléculaires et Supramoléculaires, UMR 5246 CNRS Université Claude Bernard Lyon 1 - CPE Lyon - INSA, 43 Boulevard du 11 Novembre
1918, 69622 Villeurbanne cedex, France

E-mail: gerald.monard@univ-lorraine.fr; laurent.vial@univ-lyon1.fr

Abstract

Despite of a growing use of the poly-L-lysine dendrigrafts in biomedical applications, a deeper understanding of the molecular level properties of these macromolecules

is missing. Herein, we report a simple methodology for the construction of three-dimensional structures of poly-L-lysine dendrigrafts, and the subsequent investigation of their structural features using microsecond molecular dynamics simulations. This methodology relies on the encoding of the polymers— experimental characterizations (i.e. composition, degrees of polymerization, branching ratios, charges) into alphanumeric strings that are readable by the Amber simulation package. Such an original approach opens avenues toward the in silico exploration of dendrigrafts and hyperbranched polymers.

Introduction

Dendrigrafts of poly-L-lysine (DGLs) recently complemented the family of polycationic dendritic macromolecules that include the prominent polyamidoamine (PAMAM) and polyethylenimine (PEI) polymers (Figure 1). Because their iterative synthesis in aqueous conditions is thermodynamically controlled by precipitation and kinetically controlled by steric hindrance, DGLs share features with both PAMAM dendrimers (i.e. generation-based growth and narrow molecular weight distribution), and hyperbranched PEI (i.e. broad number of regioisomers and rapid increase in molecular weight per generation). In contrast to PAMAM and PEI, DGLs are biodegradable, exhibit low cellular toxicities, and turned out to be non-immunogenic. As a consequence, DGLs have recently gained a huge interest in the biomedical field during the last quinquennium, with numerous applications in drug and gene delivery, 2,3,5-18 biomaterials and tissue engineering, 19-25 bio-imaging 26-31 and biosensing. 32-34 Despite this growing use of DGLs, a deeper understanding of the molecular level properties of these macromolecules is missing. In this context, we decided to explore the possibility to construct DGLs in silico and subsequently perform molecular dynamics (MD) simulations.

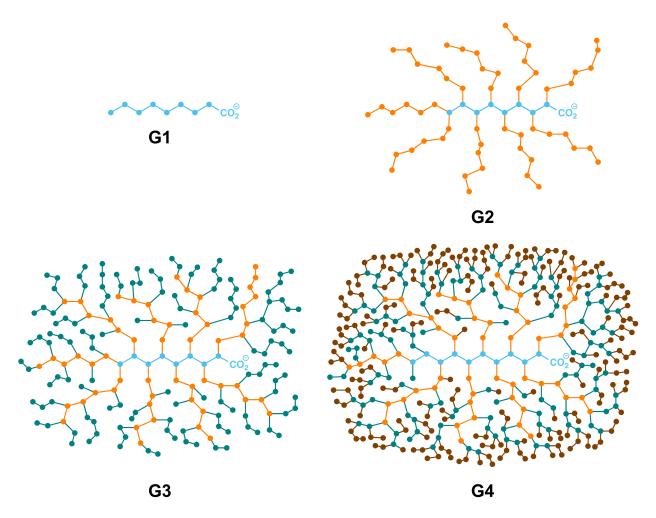


Figure 1: Schematic representation of the first- to fourth-generation DGLs $\mathbf{G1}$ - $\mathbf{G4}$ (each dot represents a L-lysine residue, pending free amino groups are not represented).

Results and Discussion

Experimental Data

As DGLs are only made of L-lysine amino acids, one could assimilate them to proteins. However, some of the residues are linked to each other through isopeptidic bonds (i.e. amide linkages that involve the nitrogen atom of the lateral chain of lysine residues). From the data that were previously collected in our group, 1 we extracted the degree of polymerization obtained from size exclusion chromatography (DP) and the mean branching ratio obtained from ¹H NMR spectroscopy (BR, defined as the ratio of the number of ϵ -branched Lysine residues over the total number of residues DP), which allowed us to calculate the number of peptidic and isopeptidic bonds, as well as the number of free α - and free ϵ -amine functions for each dendrigraft generation G1-G4 (Table 1). Also, determining the protonation state at physiological pH of each generation of DGL was obviously required in order to perform accurate modeling experiments. When a molecule bears several acidic functions, the protonation state of each of them influences the intrinsic acidity of the other ones, resulting in a shift of the functions' pKa. This phenomenon is called the polyelectrolyte effect.³⁵ Moreover, these pKa are also influenced by several other parameters, such as the solvent exposition of the acidic functions, or their capacity to form hydrogen bonds. While the pKa of the α -amine in an isolated lysine residue is 9.16, the literature reports an average value of 7.7 for proteins' N-termini, with a low value of 6.8. 36,37 We performed pH-metric titrations on DGLs (with trifluoroacetate TFA as counterions) in deionized water. The samples were fully acidified with 0.5 mL of 1M nitric acid, and then submitted to incremental additions of 0.5M sodium hydroxide using a 702 SM Titrino automatic titrator. The resulting titration curves are shown in figure 2.

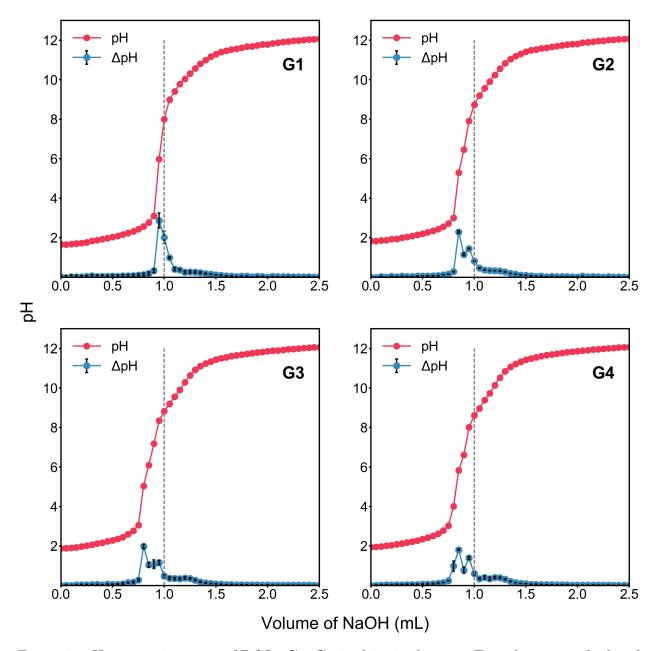


Figure 2: pH-metric titrations of DGLs **G1-G4** in deionized water. Error bars are calculated from the standard deviation of each pH measurement.

One would expect the pH to be stable below 1 mL of added NaOH since it was the required volume to neutralize the amount of nitric acid used to acidify the samples. However all the titration curves displayed a "lack" of nitric acid. This gap could be explained by the fact that not all the amine functions were protonated in the commercial powders. ³⁸ Thus, we made the hypothesis that a part of the nitric acid added in a first place was consumed by

the protonation of these neutral amino groups. The second peaks of the derivatives allowed us to calculate the number of α -amino groups in the samples, except for DGLs **G1** since the presence of a unique α -amino group did not result in a mesurable signal by potentiometry. Due to the polyelectrolyte effect, the third peaks of the derivatives – corresponding to the pKa of the ϵ -amines – were barely visible. The deprotonation of these functions takes place on a wide range of pH, making it impossible to determine the number of ϵ -amines, which was therefore calculated from the ratio $\frac{\alpha}{\epsilon}$ (see table 1). The electroneutrality equation was then used to calculate the protonation fraction Θ (equation 1), where C_a is the concentration of added acid, C_b the concentration of added base, C_{TFA} the missing concentration of TFA, C_p the total concentration of amino groups, and C_{H^+} and C_{OH^-} the concentration of free H⁺ and free OH⁻, respectively.

$$\Theta = \frac{C_a - C_b - C_{TFA} - C_{H^+} + C_{OH^-}}{C_p} \tag{1}$$

$$\Theta = \frac{C_a - C_b - C_{TFA} - 10^{-pH} + \frac{10^{-pK_w}}{10^{-pH}}}{C_p} \tag{2}$$

Equation 1 could also be expressed as a function of the pH, leading to equation 2. Applied on the experimental titration data, equation 2 gave the protonation degree of DGLs **G1-G4** at a physological pH of 7.8, which was selected since all the reported applications for DGLs are in the biomedical field (Table 2). It turned out that this percentage of non-protonated amine functions are in excellent agreement with the percentage of α -amine functions in the DGLs. This suggested that only the ϵ -amines were protonated at pH 7.8.

Table 1: Selected properties of DGLs ${f G1\text{-}G4}$. DP: degree of polymerization; BR: branching ratio.

DGL	DP	BR (%)	Free α -amino groups	Free ϵ -amino groups	Isopeptidic bonds	Peptidic bonds
G1	8	0	1	8	0	7
G2	48	12	7	42	6	41
G3	123	24	31	93	30	92
G4	365	23	85	281	84	280

Table 2: Protonation state of DGLs **G1-G4** at pH 7.8. Θ : protonation fraction; ϵ : number of free ϵ -amino groups; α : number of free α -amino groups. We assumed that, as for DGLs **G2-G4**, only the ϵ -amino groups of DGL **G1** were protonated.

DGL	Θ (%)	$\frac{\epsilon}{\epsilon + \alpha} \ (\%)$	Net charge
G1	N/A	N/A	+7
G2	86	86	+41
G3	79	75	+92
G4	79	78	+280

Encoding

From these experimental parameters, a manual approach toward an initial set of threedimensional (3D) structures for DGLs may suffer from several flaws: i) it is highly timeconsuming to manually edit starting structures that contain all atoms and residues with a correct naming in a coherent order, as well as no crossing nor bumping between the branches, ii) it is very much error-prone since the fourth-generation G4 consists of 365 residues, iii) it is likely to be biased toward an homogeneous distribution of the residues and unlikely to be representative of the experimental characteristics of the polymers. For these reasons, a new and automated method has been designed for the construction of 3D structures of DGLs. As a first step, the macromolecules G1-G4 have been deconvoluted into six discrete building blocks, which encompass all the possible arrangements (i.e. peptidic or isopeptidic bonds, charged or neutral nitrogen atoms) of the L-lysine residues within the dendrigrafts (Figure 3). Each building block was subsequently labeled with a one-letter code that unambiguously identifies it. DGL G1 could be written as the following alphanumeric string: ACCCCCCh (from C-terminal to N-terminal). It should be highlighted that DGLs were analogously synthetized in silico and at the bench, with the growth of the chains from C-terminal to the N-terminal using N ϵ -protected-L-lysine residues. Moving to a higher generation required the deprotection of the lysines-side chains, and elongation/branching/termination events from the free nitrogen atoms. 1 For the higher generation G2-G4, numbers have been introduced within the strings, which addressed the locations of the residues where branched chains were growing from (starting from the C-terminal amino acid). For instance, a possible representation of a G2 polymer using these molecular descriptors would be an alphanumeric string of 48 letters, and 6 discrete numbers (Figure 4, Top). As a result, these alphanumeric strings encoded not only the position of every L-Lysine residues in the macromolecules, but also their connections and protonation states (Figure 4, Bottom). The algorithm that built this sequence works as follows (Figure 5). Starting from a single monomer, the chain was elongated through the iterative formation of peptidic bonds until the DP (= 8) of DGL G1 was reached.

Then, 6 (= DP × BR) residues were randomly branched to **G1** in order to respect the experimental number of isopeptidic bonds. A loop was subsequently used to randomly grow the polymer through peptidic bonds until the DP (= 48) of DGL **G2** was reached. Finally, the algorithm walked through the tree and translated it into the following alphanumeric string BCDCDDDCCh8CCCCCh7CCCCCCCCCCCCh6CCCCCh5CCCCh3CCCCh1CCCCh, which encodes one possible structure of DGL **G2**. Using this algorithm, a homemade python script ³⁹ allowed us to generate eight random – but respecting the initial experimental parameters (*i.e.* DP, BR, and net charge) – sequences for each generation of DGL **G2** to **G4** (see ESI†). Within a set, each new generation of DGL was grown from the corresponding lower one.

The coded polymers could then be converted to candidate 3D structures by using a customized Nucleic Acid Builder (NAB) program ⁴⁰ from the Amber simulation package. ^{39,41} To put it simply, each one-letter coded residue was sequentially read by the program (from C-terminal to N-terminal), translated to a three-letter equivalent in order to satisfy the PDB three-letter conventional naming of the residues required by AMBER preparatory tools (see Figure 3), converted to its corresponding 3D molecular building block, and finally added at the right position to the growing dendrigraft. Preparatory *ab initio* quantum mechanics calculations (Hartree-Fock method, at the 6-31G level) were performed in order to assign atomic partial charges for the six building blocks **A-h** by using the Antechamber module of the Amber simulation Package. After each addition, the positions of all current atoms were optimized using a short conjugated gradient minimization (cut-off of 15 Å, with a maximum of 100 iterations) in order to ensure that no bond or atom overlapping was caused.

Molecular Dynamics

We performed an investigation on the structural features of DGLs by using MD simulations on the full set of 3D structures of the dendrigrafts that were generated previously (1 structure for DGL G1, 8 structures for each upper generation, 25 polymers in total). All-atom

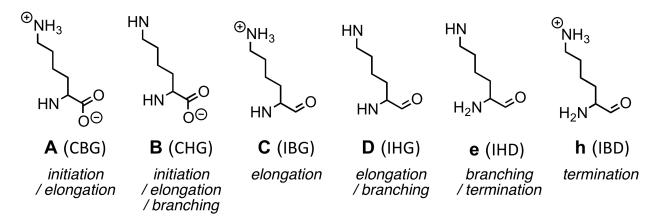


Figure 3: All possible L-Lysine buildings blocks \mathbf{A} - \mathbf{h} (and their corresponding three-letter codes in brackets) involved in the construction of DGLs $\mathbf{G1}$ - $\mathbf{G4}$.

BCDCDDDDCCh8CCCCCh7CCCCCCCCCCh6CCCCCh5CCCCh3CCCCh1CCCCh

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 C-terminal

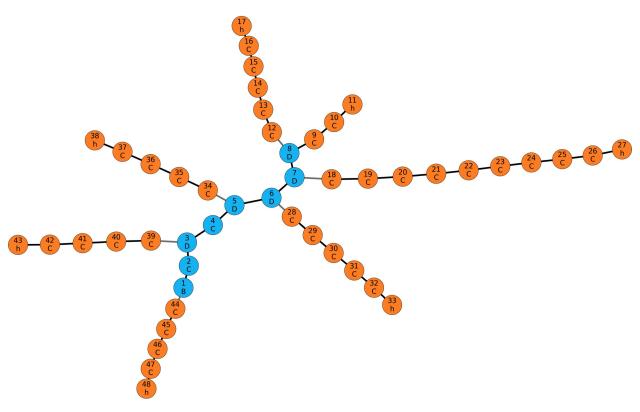


Figure 4: The alphanumeric description of a second-generation dendrigraft G2 (Top), and the corresponding arrangement of the residues (Bottom). Residues in blue indicate the mother DGL G1. Peptidic and isopeptidic bonds are depicted as black line and grey lines, respectively.

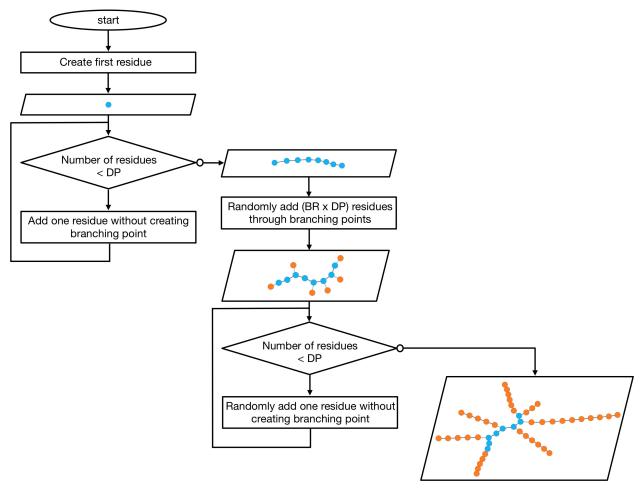


Figure 5: Flowchart toward the encoding of one possible strucure of DGL G2, which respects its experimental characteristics.

classical MD simulations were run on NVIDIA graphics processing units and included the generalized Born solvation model, which previously proved to yield the best agreement between calculated and experimental structures for proteins. ⁴² Also, the method of hydrogen mass repartitioning was applied for further accelerating the simulations. ⁴³ Using this setup, trajectories of 10 microseconds for first- and second-generation DGLs, and of 1 microsecond for third- and fourth-generation DGLs could be obtained within reasonable calculation time scales. The mean calculation times (with simulation times in parentheses) on NVIDIA K20 GPUs were 207.6 h (10 μ s) for G1, 362.6 h (10 μ s) for G2, 78.4 h (1 μ s) for G3, and 432.2 h (1 μ s) for G4. At this point, it should be noted that – albeit each dendrigraft within a same generation displays a unique topology – the analyses of the trajectories did not revealed divergent dynamic behaviours. As a consequence, the post-calculation analyses for only one dendrigraft per generation are shown in the manuscript (for the full set, see ESI†).

Via the Flory theory approach, we obtained the scaling relation $Rg \sim N^{0.41}$, where Rg is the radius of gyration and N the number of residues of the dendrigrafts. ⁴⁴ The exponent of the power law lies in between the calculated values at physiological pH of 0.34 and 0.57 for poly-L-lysine dendrimers and linear poly-L-lysines, respectively. ^{45–47} This suggests that the dendrigrafts adopt intermediate morphologies, which may remind those of unfolded and intrinsically disordered proteins. ⁴⁸ This conformational flexibility was further supported by the 2D root-mean-square deviations of selected residues ⁴⁹ (see ESI†) and the Ramachandran plots of all possible residues (data not shown) along the MD trajectories, from which it was clear that no folded structure was clearly identifiable for any of the 4 generations of dendrigraft **G1-G4**.

In order to assess the shape of the dendrigrafts, the asphericity parameter (Δ) and shape parameter (S) can be evaluated from the inertial tensor⁵⁰ described in equation 3, where M is the total mass, m_i and r_i are the mass and position of the ith residue, and $\alpha, \beta = x, y, z$.

$$T_{\alpha,\beta} = \frac{1}{2M^2} \sum_{j=1}^{N} \sum_{j=1}^{N} m_i m_j (r_{i\alpha} - r_{j\alpha}) (r_{i\beta} - r_{j\beta})$$
(3)

The squares of the three principal radii of gyration— R_G —are calculated from the trace of the matrix— $T(\alpha, \beta)$, which amounts to the sum of its eigenvalues— λ_1 ,— λ_2 and— λ_3 (equation 4).

$$R_G^2 = trT = \sum_{i=1}^3 \lambda_i \tag{4}$$

These eigenvalues are used to calculate Δ , which measures the average deviation from spherical symmetry (equation 5).

$$\Delta = \frac{3}{2} \frac{\sum_{i=1}^{3} (\lambda_i - \lambda)^2}{(trT)^2} \tag{5}$$

Where

$$\lambda = \frac{\sum_{i=1}^{3} \lambda_i}{3} \tag{6}$$

On the other hand, S further describes the ellipticities of the polymers. In three dimensions, S can be calculated from equation 7.

$$S = 27 \frac{\prod_{i=1}^{3} (\lambda_i - \lambda)}{(trT)^3} \tag{7}$$

When applied to our trajectories, these treatments gave almost null Δ and S values (see ESI† for full set of values). While the theoretical lowest value of Δ is zero (i.e. characterizing a sphere) and the highest value is one (i.e. characterizing a rod), the values of S are theoretically bounded following the inequality $-\frac{1}{4} \leq S \leq 2$, with negative and positive values indicating oblate and prolate objects, respectively. When applied to our trajectories, the calculated values actually establish that our dendrigrafts are mostly globular, as they previously were represented in the literature (Table 3), with the exception of the first-generation G1 that slightly tends toward a disc-shaped rod due to its linear topology. Insights into the spatial occupancy of the macromolecules were gained by combining the atomic coordinates

of each frame – captured every nanosecond for **G1** and **G2**, and every 100 picoseconds for **G3** and **G4** – over the entire MD trajectory. The resulting volumetric maps showed that the DGLs tend to occupy a large amount of the available space as a result of the flexibility of their branches (Figure 6).

Table 3: Averages and standard deviations of the asphericity (Δ) and shape (S) parameters for first- to fourth-generation DGLs **G1** to **G4**.

	G1	G2	G3	G4
Δ 0.107 :	$\pm 0.034 \ 0.02$	9 ± 0.014 (0.023 ± 0.007	0.008 ± 0.002
S -0.069	± 0.032 -0.00	0.006 - 0.006	-0.004 ± 0.003	-0.001 ± 0.001

Then, we plotted both the normalized atomic and cationic densities as a function of their distances from the center of mass of the polymers (Figure 7). Such an analysis revealed that: i) the atomic density shifts from the core toward the outside of the dendrigrafts with increasing generation, and ii) the cationic charges are mainly found on the outer shell layer of the dendrigrafts. The distribution of the positive charges, as well as the bulk density of the dendrigrafts, can also be appreciated from snapshots taken after 1 μ s of the trajectories (Figure 8). Finally, we measured the distance between all adjacent charged nitrogen atoms (NZ) using the pair correlation function (Figure 9). It turned out that all DGLs – independently to their generation – displayed adjacent NZ separated by a similar average distance of 8.0 Å (first peak of the curves). This latter observation relates to the synthetic route towards DGLs, which implies a N $^{\epsilon}$ -protected lysine derivative and therefore excludes an impact of the electrostatic repulsion on the ionizable nitrogen atoms' distribution.

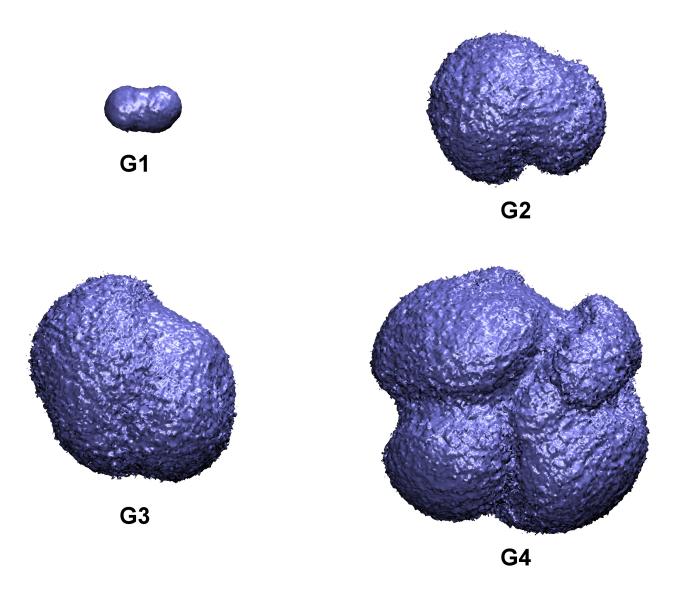


Figure 6: Volumetric maps of first- to fourth-generation DGLs ${\bf G1\text{-}G4}$ at 1 microsecond with 99% probability boundary surfaces. All images are to the same scale. Images were generated with VMD (ref. 52).

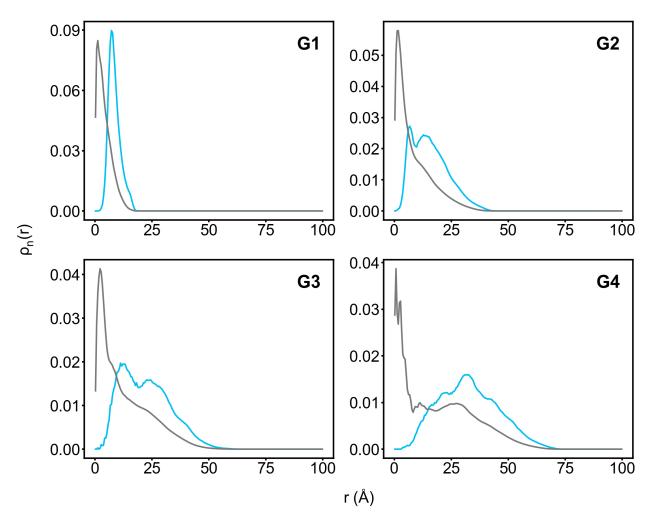


Figure 7: Normalized atomic ($in\ grey$) and cationic ($in\ blue$) density profiles of first- to fourth-generation DGLs **G1-G4** (ref. 51).

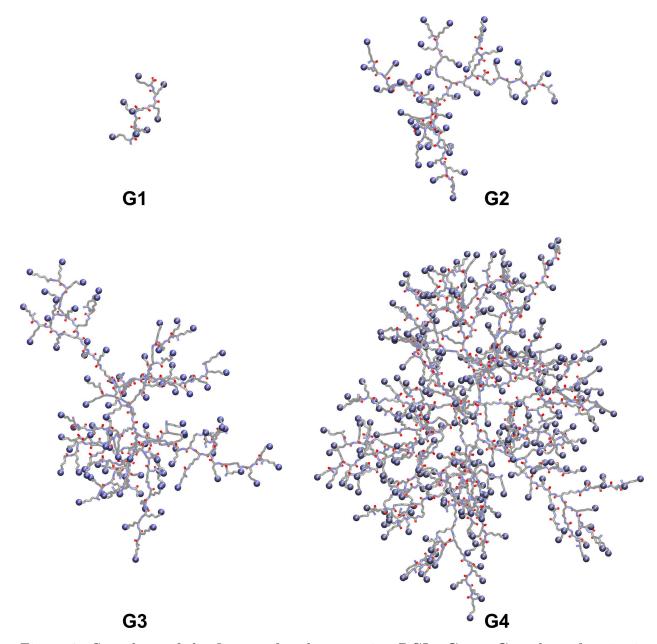


Figure 8: Snapshots of the first- to fourth-generation DGLs ${\bf G1}$ to ${\bf G4}$ taken after 1 microsecond at T = 300 K. Blue dots highlight the positively charged nitrogen atoms at pH 7.8. All images are to the same scale.

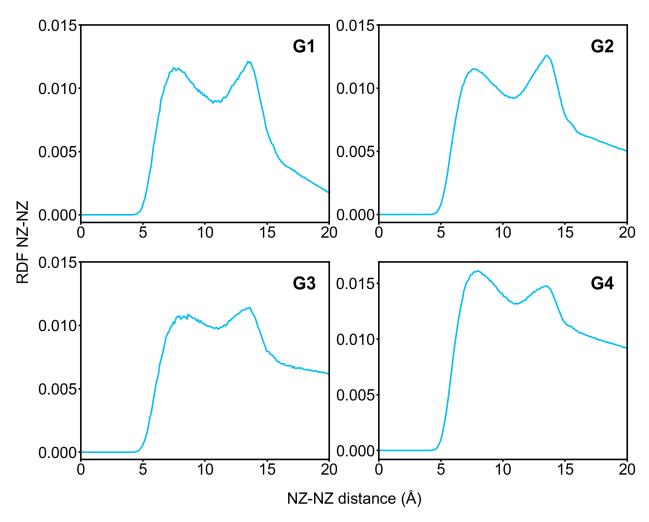


Figure 9: Radial Distribution Function (RDF) of the charged nitrogen atoms (NZ) for the first- to fourth-generation $\mathbf{G1}$ to $\mathbf{G4}$.

By being able to generate accurate 3D structures of four generations of poly-L-lysines dendrigrafts with respect to their experimental characteristics (*i.e.* composition, degrees of polymerization, branching ratios, number of charges at physiological pH), and to determine their molecular properties (*i.e.* topology, shape, flexibility), one should now be able to study at the molecular level their interactions with biologically relevant partners such as drugs, metabolites, biopolymers, genetic materials, proteins, or cells. These computational studies may not only provide important information to apprehend their unique features such as biocompatibility, 3,4,12 membrane translocation ability, 12 and highly efficient electrostatic binding property, 32 but consequently also lead to the development of DGLs for real-world

applications in nanomedicine.

Computational Details

Residues were prepared with the antechamber module of AmberTools. ⁴¹ Atomic charges were generated from RESP charges extracted from HF/6-31G* optimized geometries of capped residues. Initial structures were built using the Nucleic Acid Builder (NAB) program ⁴⁰ and the LEaP module of AmberTools. From the primary sequence of one dendrimer, new residues were added one by one and, at each new addition, a minimization of the full system was performed for 100 steps. All molecular dynamics simulations were carried out using the GPU implementation ⁵³ of the pmemd program in Amber15 ⁴¹ with the combination of GB-Neck2 model, ⁵⁴ mbondi3 radii, ⁵⁴ and ff14SB. ⁵⁵ No cutoff was applied for computing the nonbonded interactions. A time step of 4 fs was used with hydrogen mass repartitioning. ⁵⁶ Bonds involving hydrogen were constrained by the SHAKE algorithm. ⁵⁷ Temperature was controlled with a Langevin thermostat at 300 K with a collision frequency $\gamma = 1.0 \text{ ps}^{-1}$.

Conclusions

In conclusion, microsecond molecular dynamics simulations on second- to fourth-generation poly-L-lysine dendrigrafts revealed a myriad of conformational states accessible to them without folded states. This conformational plasticity (*i.e.* this ability to adaptably display positive charges to make ion pairs) may explain why DGLs are such efficient binders for anionic ligands even in the most competitive media. In addition, the generated structures can now be exploited for in-depth analyses of the interactions between these macromolecules and biologically relevant partners for future biomedical applications. More broadly, this work provides to the community a general approach toward the construction of dendrigrafts – but also hyperbranched polymers – $in\ silico$. We are therefore confident that molecular dynamics will become a pervasive tool to explore the structural features of large dendrigrafts

and hyperbranched polymers at the molecular level.

Acknowledgement

This work was supported by the Centre National de la Recherche Scientifique (CNRS), the Université de Lorraine, and the Université de Montpellier. GPU resources were allocated by the Centre de Calcul ROMEO of the Université de Reims Champagne-Ardenne. DGLs were supplied by the COLCOM company (Montpellier, France).

Supporting Information Available

The following files are available free of charge.

Alphanumeric strings, 2D root-mean-square deviations, asphericity and shape parameters, normalized atomic and cationic densities, radial distribution function, and volumetric maps of all polymers (PDF).

References

- (1) Collet, H.; Souaid, E.; Cottet, H.; Deratani, A.; Boiteau, L.; Dessalces, G.; Rossi, J.-C.; Commeyras, A.; Pascal, R. An Expeditious Multigram-Scale Synthesis of Lysine Dendrigraft (DGL) Polymers by Aqueous N-Carboxyanhydride Polycondensation. *Chem. Eur. J.* 2010, 16, 2309–2316.
- (2) Kodama, Y.; Nakamura, T.; Kurosaki, T.; Egashira, K.; Mine, T.; Nakagawa, H.; Muro, T.; Kitahara, T.; Higuchi, N.; Sasaki, H. Biodegradable Nanoparticles Composed of Dendrigraft Poly-L-Lysine for Gene Delivery. Eur. J. Pharm. and Biopharm. 2014, 87, 472–479.
- (3) Tang, M.; Dong, H.; Li, Y.; Ren, T. Harnessing the PEG-Cleavable Strategy to Balance

- Cytotoxicity, Intracellular Release and the Therapeutic Effect of Dendrigraft Poly-L-Lysine for Cancer Gene Therapy. J. Mater. Chem. B 2016, 4, 1284–1295.
- (4) Romestand, B.; Rolland, J.-L.; Commeyras, A.; Coussot, G.; Desvignes, I.; Pascal, R.; Vandenabeele-Trambouze, O. Dendrigraft Poly-L-Lysine: A Non-Immunogenic Synthetic Carrier for Antibody Production. *Biomacromolecules* 2010, 11, 1169–1173.
- (5) Liu, Y.; An, S.; Li, J.; Kuang, Y.; He, X.; Guo, Y.; Ma, H.; Zhang, Y.; Ji, B.; Jiang, C. Brain-targeted Co-Delivery of Therapeutic Gene and Peptide by Multifunctional Nanoparticles in Alzheimer's Disease Mice. *Biomaterials* 2016, 80, 33–45.
- (6) Yang, Z.-Z.; Li, L.; Wang, L.; Xu, M.-C.; An, S.; Jiang, C.; Gu, J.-K.; Wang, Z.-J. J.; Yu, L.-S.; Zeng, S. siRNA Capsulated Brain-targeted Nanoparticles Specifically Knock Down OATP2B1 in Mice: a Mechanism for Acute Morphine Tolerance Suppression. Sci. Rep. 2016, 6, 33338.
- (7) Kodama, Y.; Kuramoto, H.; Mieda, Y.; Muro, T.; Nakagawa, H.; Kurosaki, T.; Sakaguchi, M.; Nakamura, T.; Kitahara, T.; Sasaki, H. Application of Biodegradable Dendrigraft Poly-L-Lysine to a Small Interfering RNA Delivery System. J. Drug Target. 2016, 1–9.
- (8) Yao, H.; Wang, K.; Wang, Y.; Wang, S.; Li, J.; Lou, J.; Ye, L.; Yan, X.; Lu, W.; Huang, R. Enhanced Blood-brain Barrier Penetration and Glioma Therapy Mediated by a New Peptide Modified Gene Delivery System. *Biomaterials* **2015**, *37*, 345–352.
- (9) Han, L.; Guo, Y.; Ma, H.; He, X.; Kuang, Y.; Zhang, N.; Lim, E.; Zhou, W.; Jiang, C. Acid Active Receptor-Specific Peptide Ligand for In Vivo Tumor-Targeted Delivery. Small 2013, 9, 3647–3658.
- (10) An, S.; Kuang, Y.; Shen, T.; Li, J.; Ma, H.; Guo, Y.; He, X.; Jiang, C. Brain-targeting Delivery for RNAi Neuroprotection Against Cerebral Ischemia Reperfusion Injury. *Bio-materials* 2013, 34, 8949–8959.

- (11) Liu, Y.; Guo, Y.; An, S.; Kuang, Y.; He, X.; Ma, H.; Li, J.; Lv, J.; Zhang, N.; Jiang, C. Targeting Caspase-3 as Dual Therapeutic Benefits by RNAi Facilitating Brain-Targeted Nanoparticles in a Rat Model of Parkinson's Disease. *PLoS ONE* **2013**, *8*, e62905.
- (12) Hofman, J.; Buncek, M.; Haluza, R.; Streinz, L.; Ledvina, M.; Cigler, P. In Vitro Transfection Mediated by Dendrigraft Poly-*L*-lysines: The Effect of Structure and Molecule Size. *Macromol. Biosci.* **2013**, *13*, 167–176.
- (13) Liu, Y.; Li, J.; Shao, K.; Huang, R.; Ye, L.; Lou, J.; Jiang, C. A Leptin Derived 30-Amino-Acid Peptide Modified Pegylated Poly-*L*-Lysine Dendrigraft for Brain Targeted Gene Delivery. *Biomaterials* **2010**, *31*, 5246–5257.
- (14) Huang, R.; Liu, S.; Shao, K.; Han, L.; Ke, W.; Liu, Y.; Li, J.; Huang, S.; Jiang, C. Evaluation and Mechanism Studies of PEGylated Dendrigraft Poly-L-Lysines as Novel Gene Delivery Vectors. Nanotechnology 2010, 21, 265101.
- (15) Hu, G.; Zhang, H.; Zhang, L.; Ruan, S.; He, Q.; Gao, H. Integrin-mediated Active Tumor Targeting and Tumor Microenvironment Response Dendrimer-gelatin Nanoparticles for Drug Delivery and Tumor Treatment. *Int. J. Pharm.* **2015**, *496*, 1057–1068.
- (16) Huang, S.; Shao, K.; Liu, Y.; Kuang, Y.; Li, J.; An, S.; Guo, Y.; Ma, H.; Jiang, C. Tumor-Targeting and Microenvironment-Responsive Smart Nanoparticles for Combination Therapy of Antiangiogenesis and Apoptosis. ACS Nano 2013, 7, 2860–2871.
- (17) Liu, S.; Guo, Y.; Huang, R.; Li, J.; Huang, S.; Kuang, Y.; Han, L.; Jiang, C. Gene and Doxorubicin Co-Delivery System for Targeting Therapy of Glioma. *Biomaterials* **2012**, 33, 4907–4916.
- (18) Sideratou, Z.; Sterioti, N.; Tsiourvas, D.; Tziveleka, L.-A.; Thanassoulas, A.; Nounesis, G.; Paleos, C. M. Arginine End-functionalized Poly-L-Lysine) Dendrigrafts for the Stabilization and Controlled Release of Insulin. J. Colloid Interface Sci. 2010, 351, 433–441.

- (19) Kojima, C.; Fusaoka-Nishioka, E.; Imai, T.; Nakahira, A.; Onodera, H. Dendrigraft Polylysine Coated-poly(glycolic acid) Fibrous Scaffolds for Hippocampal Neurons. J. Biomed. Mater. Res. A 2016, 104, 2744–2750.
- (20) Lorion, C.; Faye, C.; Maret, B.; Trimaille, T.; Régnier, T.; Sommer, P.; Debret, R. Biosynthetic Support Based on Dendritic Poly-L-Lysine Improves Human Skin Fibroblasts Attachment. J. Biomater. Sci. Polym. Ed. 2014, 25, 136–149.
- (21) Eap, S.; Bécavin, T.; Keller, L.; Kokten, T.; Fioretti, F.; Weickert, J.-L.; Deveaux, E.; Benkirane-Jessel, N.; Kuchler-Bopp, S. Nanofibers Implant Functionalized by Neural Growth Factor as a Strategy to Innervate a Bioengineered Tooth. Adv. Healthc. Mater. 2014, 3, 386–391.
- (22) Eap, S.; Ferrand, A.; Schiavi, J.; Keller, L.; Kokten, T.; Fioretti, F.; Mainard, D.; Ladam, G.; Benkirane-Jessel, N. Collagen Implants Equipped with 'Fish Scale'-like Nanoreservoirs of Growth Factors for Bone Regeneration. *Nanomedicine* 2014, 9, 1253– 1261.
- (23) Cadiere, A.; Couturaud, B.; Boismard, J.; Cann, P. L.; Gérard, A.; Mas, A.; Faye, C.; Garrelly, L.; Roig, B. Assessment of Poly-L-Lysine Dendrigrafts for Virus Concentration in Water: Use of MS2 Bacteriophage as Proof of Concept. J. Appl. Microbiol. 2013, 115, 290–297.
- (24) Mendoza-Palomares, C.; Ferrand, A.; Facca, S.; Fioretti, F.; Ladam, G.; Kuchler-Bopp, S.; Regnier, T.; Mainard, D.; Benkirane-Jessel, N. Smart Hybrid Materials Equipped by Nanoreservoirs of Therapeutics. ACS Nano 2012, 6, 483–490.
- (25) Fioretti, F.; Mendoza-Palomares, C.; Helms, M.; Alam, D. A.; Richert, L.; Arntz, Y.; Rinckenbach, S.; Garnier, F.; Haikel, Y.; Gangloff, S. C.; Benkirane-Jessel, N. Nanostructured Assemblies for Dental Application. *ACS Nano* **2010**, *4*, 3277–3287.

- (26) Han, L.; Liu, M.; Ye, D.; Zhang, N.; Lim, E.; Lu, J.; Jiang, C. Tumor Cell Membrane-targeting pH-dependent Electron Donor-Acceptor Fluorescence Systems with Low Background Signals. *Biomaterials* **2014**, *35*, 2952–2960.
- (27) Li, J.; Huang, S.; Shao, K.; Liu, Y.; An, S.; Kuang, Y.; Guo, Y.; Ma, H.; Wang, X.; Jiang, C. A Choline Derivate-modified Nanoprobe for Glioma Diagnosis using MRI. Sci. Rep. 2013, 3, 1623–1632.
- (28) Liu, Y.; Hu, Y.; Guo, Y.; Ma, H.; Li, J.; Jiang, C. Targeted Imaging of Activated Caspase-3 in the Central Nervous System by a Dual Functional Nano-Device. *J. Control. Release* **2012**, *163*, 203–210.
- (29) Huang, R.; Han, L.; Li, J.; Liu, S.; Shao, K.; Kuang, Y.; Hu, X.; Wang, X.; Lei, H.; Jiang, C. Chlorotoxin-modified Macromolecular Contrast Agent for MRI Tumor Diagnosis. *Biomaterials* **2011**, *32*, 5177–5186.
- (30) Theodossiou, T. A.; Sideratou, Z.; Tsiourvas, D.; Paleos, C. M. A Novel Mitotropic Oligolysine Nanocarrier: Targeted Delivery of Covalently Bound D-Luciferin to Cell Mitochondria. *Mitochondrion* 2011, 11, 982–986.
- (31) Kojima, C.; Turkbey, B.; Ogawa, M.; Bernardo, M.; Regino, C. A.; Bryant, L. H.; Choyke, P. L.; Kono, K.; Kobayashi, H. Dendrimer-based MRI Contrast Agents: The Effects of PEGylation on Relaxivity and Pharmacokinetics. *Nanomedicine: NBM* 2011, 7, 1001–1008.
- (32) Francoia, J.-P.; Pascal, R.; Vial, L. Monitoring Clinical Levels of Heparin in Human Blood Samples with an Indicator-Displacement Assay. Chem. Commun. 2015, 51, 1953–1956.
- (33) Francoia, J.-P.; Vial, L. A KISS (keep it simple, sensor) Array for Glycosaminoglycans. *Chem. Commun.* **2015**, *51*, 17544–17547.

- (34) Couturaud, B.; Bondia, A. M.; Faye, C.; Garrelly, L.; Mas, A.; Robin, J. J. Grafting of Poly-L-Lysine Dendrigrafts onto Polypropylene Surface Using Plasma Activation for ATP Immobilization Nanomaterial for Potential Applications in Biotechnology. J. Colloid and Interface Sci. 2013, 408, 242–251.
- (35) Koper, G. J. M.; Borkovec, M. Proton Binding by Linear, Branched, and Hyperbranched Polyelectrolytes. *Polymer* **2010**, *51*, 5649–5662.
- (36) Thurlkill, R. L.; Thurlkill, R. L.; Grimsley, G. R.; Grimsley, G. R.; Scholtz, J. M.; Scholtz, J. M.; Pace, C. N.; Pace, C. N. pK Values of the Ionizable Groups of Proteins. Protein Sci. 2006, 15, 1214–1218.
- (37) Pace, C. N.; Grimsley, G. R.; Scholtz, J. M. Protein Ionizable Groups: pK Values and their Contribution to Protein Stability and Solubility. *J. Biol. Chem.* **2009**, *284*, 13285–13289.
- (38) DGLs were provided by the Colcom company http://www.colcom.eu (accessed September 2017).
- (39) The source codes of the programs are available at https://www.github.com/gmonard/DGL (accessed January 2017).
- (40) Macke, T. J.; Case, D. A. In Molecular Modeling of Nucleic Acids; Leontis, N. B., SantaLucia, J., Eds.; American Chemical Society: Washington, DC, 1997; Chapter 24, pp 379–393.
- (41) Case, D. A.; Berryman, J. T.; Betz, R. M.; Cerutti, D. S.; Cheatham, T. E.; Darden, T. A.; Duke, R. E.; Giese, T. J.; Gohlke, H.; Goetz, A. W.; Homeyer, N.; Izadi, S.; Janowski, P.; Kaus, J.; Kovalenko, A.; Lee, T. S.; LeGrand, S.; Li, P.; Luchko, T.; Luo, R.; Madej, B.; Merz, K. M.; Monard, G.; Needham, P.; Nguyen, H.; Nguyen, H. T.; Omelyan, I.; Onufriev, A.; Roe, D. R.; Roitberg, A.; Salomon-Ferrer, R.; Simmerling, C. L.; Smith, W.; Swails, J.; Walker, R. C.; Wang, J.; Wolf, R. M.; Wu, X.;

- York, D. M.; Kollman, P. A. AMBER 15: University of California, San Francisco, 2015.
- (42) Nguyen, H.; Maier, J.; Huang, H.; Perrone, V.; Simmerling, C. Folding Simulations for Proteins with Diverse Topologies Are Accessible in Days with a Physics-Based Force Field and Implicit Solvent. J. Am. Chem. Soc 2014, 136, 13959–13962.
- (43) Hopkins, C. W.; Grand, S. L.; Walker, R. C.; Roitberg, A. E. Long-Time-Step Molecular Dynamics through Hydrogen Mass Repartitioning. *J. Chem. Theory Comput.* **2015**, *11*, 1864–1874.
- (44) Flory, P. J. The Configuration of Real Polymer Chains. J. Chem. Phys. 1949, 17, 303–310.
- (45) Falkovich, S.; Markelov, D.; Neelov, I.; Darinskii, A. Are Structural Properties of Dendrimers Sensitive to the Symmetry of Branching? Computer Simulation of Lysine Dendrimers. *Journal of Chemical Physics* 2013, 139.
- (46) Jin, X.; Leclercq, L.; Sisavath, N.; Cottet, H. Investigating the Influence of Phosphate Ions on Poly(*L*-lysine) Conformations by Taylor Dispersion Analysis. *Macromolecules* 2014, 47, 5320–5327.
- (47) The theoretical values of the Flory exponent are 0.33 and 0.59 for hard spheres (*i.e.* compact spherical molecules) and random coils, respectively.
- (48) Hofmann, H.; Soranno, A.; Borgia, A.; Gast, K.; Nettels, D.; Schuler, B. Polymer Scaling Laws of Unfolded and Intrinsically Disordered Proteins Quantified with Single-Molecule Spectroscopy. *Proceedings of the National Academy of Sciences* 2012, 109, 16155–16160.
- (49) For all the polymers, only the deviations of the residues of the core DGL G1 were

- analyzed during the trajectories. The behavior of these amino acids within a polymer was considered to be representative of the whole macromolecule's behavior.
- (50) Rawdon, E. J.; Kern, J. C.; Piatek, M.; Plunkett, P.; Stasiak, A.; Millett, K. C. Effect of Knotting on the Shape of Polymers. *Macromolecules* **2008**, *41*, 8281–8287.
- (51) The normalized radial atomic (or cationic) density $\rho_n(\mathbf{r})$ was obtained by counting the number N of atoms (or positive charges) whose centers of mass are located within a spherical shell of radius r and thickness $\Delta \mathbf{r}$. The function N(r) was then divided by the volume of the shell V(r) and multiplied by a constant Θ : $\rho_n(\mathbf{r}) = \mathrm{N}(\mathbf{r}) \times \Theta / \mathrm{V}(\mathbf{r}) = \rho(\mathbf{r}) \times \Theta$, with $\Theta \times \int_0^\infty \rho(r) dr = 1$.
- (52) Humphrey, W.; Dalke, A.; Schulten, K. VMD: Visual Molecular Dynamics. J. Mol. Graph. 1996, 14, 33–38.
- (53) Goetz, A. W.; Williamson, M. J.; Xu, D.; Poole, D.; Le Grand, S.; Walker, R. C. Routine Microsecond Molecular Dynamics Simulations with AMBER on GPUs. 1. Generalized Born. J. Chem. Theory Comput. 2012, 8, 1542–1555.
- (54) Nguyen, H.; Roe, D. R.; Simmerling, C. Improved Generalized Born Solvent Model Parameters for Protein Simulations. *J. Chem. Theory Comput.* **2013**, *9*, 2020–2034.
- (55) Maier, J. A.; Martinez, C.; Kasavajhala, K.; Wickstrom, L.; Hauser, K. E.; Simmerling, C. ff14SB: Improving The Accuracy of Protein Side Chain and Backbone Parameters from ff99SB. J. Chem. Theory Comput. 2015, 11, 3696–3713.
- (56) Hopkins, C. W.; Le Grand, S.; Walker, R. C.; Roitberg, A. E. Long-Time-Step Molecular Dynamics through Hydrogen Mass Repartitioning. J. Chem. Theory Comput. 2015, 11, 1864–1874.
- (57) Ryckaert, J.-P.; Ciccotti, G.; Berendsen, H. J. Numerical Integration of the Cartesian

Equations of Motion of a System with Constraints: Molecular Dynamics of n-alkanes.

J. Comput. Phys. 1977, 23, 327–341.

Graphical TOC Entry

