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ABSTRACT

The application of terahertz radiation has been shown to affect both protein structure and cellular function. As the key to such structural changes lies in the dynamic response of a protein, we focus on the susceptibility of the protein's internal dynamics to mechanical stress induced by acoustic pressure waves. We use the open-boundary molecular dynamics method, which allows us to simulate the protein exposed to acoustic waves. By analyzing the dynamic fluctuations of the protein subunits, we demonstrate that the protein is highly susceptible to acoustic waves with frequencies corresponding to those of the internal protein vibrations. This is confirmed by changes in the compactness of the protein structure. As the amplitude of the pressure wave increases, even larger deviations from average positions and larger changes in protein compactness are observed. Furthermore, performing the mode-projection analysis, we show that the breathing-like character of collective modes is enhanced at frequencies corresponding to those used to excite the protein.

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I. INTRODUCTION

Proteins are an essential part of complex machinery that drives biological processes in living organisms. According to the sequence-structure-function paradigm raised by Anfinsen,¹ the function of a protein is governed by the unique three-dimensional structure adopted by the protein on the basis of its amino-acid sequence.^{1,2} However, in order to understand how proteins work, it is also necessary to shed light on their behavior from a dynamic point of view.³⁻⁶ Proteins at physiological temperatures exhibit a wide range of intrinsic motions,7 which span over different spatial and temporal scales and are governed by the (highly multidimensional) energy landscape.^{4,8} Conformational transitions of proteins from one substate to another, separated by high energy barriers, take place at the microsecond timescale (or even slower) and are associated with the interconversion between open and closed conformations, enzyme catalysis, signal transduction, and protein-protein interaction. Looking at a more local scale, proteins exhibit faster dynamics on the picosecond-to-nanosecond timescale, which is related to local atomic fluctuations (e.g., side chain rotations) and collective fluctuations of small groups of atoms (e.g., loop motions).^{4,9} In addition, these local protein motions have been shown to be involved in the initiation and modulation of slower dynamic

motions, such as those associated with protein conformational changes. $^{10\mathac{-12}}$

Collective motions associated with protein function can (often) be adequately described by (only) a few low-frequency normal modes.^{13–17} To computationally access functionally relevant collective motions of biomolecules, normal mode analysis (NMA) is widely used. NMA is based on the harmonic approximation of potential energy^{18,19} and assumes that large-scale motions relevant to protein function are described by normal modes with the lowest frequencies.²⁰ However, proteins at physiological temperatures exhibit anharmonicity in their motion. To this end, the principal component analysis (PCA), also implemented as the quasi-harmonic analysis (QHA),²¹ is carried out on a molecular dynamics trajectory covering full range of motions.^{22–26} Employing PCA, collective modes of internal motion are obtained as eigenvectors of the covariance matrix of atomic fluctuations.²⁷ The eigenvalues of the covariance matrix represent the variance along each of the corresponding collective modes. In PCA, only a limited number of atoms, i.e., C⁴ atoms, or selected groups of atoms can be used for analysis.²²

Both computational methods (i.e., NMA and PCA) serve as an important tool for the interpretation of spectra obtained, for example, by terahertz (THz) time-domain spectroscopy,²⁸ Raman spectroscopy,^{29,30} and neutron scattering.³¹⁻³⁵ As these experimental techniques allow access to the picosecond-to-nanosecond timescale, they are particularly suitable for detecting, identifying, and examining functionally relevant motions in biological molecules,^{30,33,36–46} as the underlying modes of collective dynamics vibrate in the THz region, which spans from 0.1 to 10 THz (or expressed in wavenumbers from 3.33 to 333 cm⁻¹).^{47–49} Moreover, THz radiation can be used for activating the process without causing denaturation or aggregation of the protein, as shown for the actin solution during polymerization.⁵⁰ In addition, THz radiation with several orders of magnitude higher peak power was displayed to cause changes in the morphology of actin filaments that were attributed to the THz-light-induced acoustic waves,^{51,52} whose generation in water was also confirmed.⁵³

THz excitations can thus be used as a tool to modify the structure of the biomolecule or to manipulate cellular functions by directly affecting protein dynamics.^{50,51,54} As illustrated, these are low-frequency internal vibrations that are key to understanding and predicting the collective dynamic behavior^{16,17,55} and should be targeted by external perturbations. To investigate protein dynamics under mechanical stress induced by acoustic waves, we select ubiquitin as the benchmark protein and model it using the coarse-grained (CG) Martini 3 model.⁵⁶ We conduct nonequilibrium simulations using the open-boundary molecular dynamics (OBMD) method^{57–60} that enables imposition of external boundary conditions (e.g., acoustic wave) by an additional external force.

II. THEORETICAL BACKGROUND AND METHODOLOGY

To inspect the protein response to induced mechanical stress, the CG protein (shown by the gray colored beads in Fig. 1) is placed at the center of the simulation box and surrounded by CG water (represented by the blue colored beads in Fig. 1). Employing OBMD, on one side of the simulation domain (i.e., the buffer region indicated by the orange rectangle in Fig. 1), sub-THz acoustic waves are generated, while on the other side (i.e., the buffer region depicted by the yellow rectangle in Fig. 1), only the constant normal load is applied.

Next, we present details on the imposition of external boundary conditions (i.e., the constant normal load and a mechanical pressure wave) using OBMD and on the computation of collective modes of the protein employing PCA.



FIG. 1. Schematic representation of the CG backbone particles of ubiquitin immersed in CG water and subjected to an acoustic wave. The region of interest (ROI) is embedded with two buffer regions (designated by the orange and yellow rectangles). The buffers serve as particle reservoirs and are crucial for the introduction of external boundary conditions on the ROI.

A. Open-boundary molecular dynamics (OBMD)

To simulate acoustic waves, we resort to OBMD,^{57–60} which has already been applied in particle-based ultrasound simulations.^{61,62} OBMD opens the molecular system, which is able to exchange mass and momentum with its surroundings.⁶³ It allows the imposition of external boundary conditions without altering Newton's equations of motion in the bulk. The simulation box is opened in one direction, i.e., in the x-direction, while in the remaining directions, i.e., y- and z-directions, periodic boundary conditions are implemented (see Fig. 1). The external boundary condition of the constant normal load p is imposed on the region of interest (ROI) from both sides of the simulation domain (in Fig. 1 depicted by the black arrows), whereas the acoustic wave is introduced from only one side (in Fig. 1 shown by the orange rectangle) as an additional oscillatory pressure contribution to the normal load p, expressed as $p + \Delta p \sin(2\pi vt)$, where Δp , v, and t are the pressure amplitude, frequency, and time, respectively.

The external boundary condition is imposed on the system through an additional external force \mathbf{F}^{ext} , acting only on particles in the buffer regions,

$$\mathbf{F}^{ext} = A \bigg(\mathbf{J}^P \cdot \mathbf{n} - \frac{\sum_{i'} \Delta(m_{i'} \mathbf{v}_{i'})}{A \delta t} \bigg), \tag{1}$$

where indices *i* and *i'* run over all particles in buffers and over all particles that have been inserted into or deleted from the system in the last time step δt , respectively. Accordingly, the momentum change is given by $\Delta(m_i \mathbf{v}_{i'}) = \pm m_{i'} \mathbf{v}_{i'}$ if the particle *i'* is deleted (–) or inserted (+), where $m_{i'}$ and $\mathbf{v}_{i'}$ stand for the mass and velocity of the particle *i'*, respectively. A is the area of the interface buffer-ROI, while **n** is the unit vector normal to the interface buffer-ROI, pointing toward the center of the simulation box. The components of the momentum flux tensor \Im^P are $J_{ij}^P = p\delta_{ij}$ for the side where only constant normal load is imposed. For the side where acoustic wave is introduced, $J_{ij}^P = (p + \Delta p \sin (2\pi v t))\delta_{ij}$. Finally, \mathbf{F}^{ext} is distributed among particles in the buffers.⁶²

The dissipative particle dynamics (DPD) thermostat is applied, as its equations conserve linear momentum and correctly reproduce the hydrodynamic behavior. $^{64-66}$

B. Principal component analysis (PCA)

Performing out-of-equilibrium OBMD simulations of the propagating acoustic wave through the protein system in water allows us to obtain an insight into variations of the protein's internal dynamics.

The optimal tool for frequency decomposition of dynamic fluctuations in the system, which, in principle, occur around a single potential energy minimum, is the PCA, also called quasi-harmonic analysis. PCA is a mathematical technique used also to reduce the number of degrees of freedom needed to describe the protein dynamics.^{27,67} It assumes that collective modes with the largest fluctuations, i.e., lowest quasi-harmonic frequencies, govern functional motions.⁶⁸ To perform PCA, the overall translation and rotation must be eliminated by superposing ensemble conformation onto a reference structure, followed by solving the eigenvalue problem of the mass-weighted variance–covariance matrix ${\sf A}.$ The components of the matrix ${\sf A}$ are

$$A_{\alpha\beta} = \sqrt{m_{\alpha}m_{\beta}} \langle (\mathbf{r}_{\alpha} - \langle \mathbf{r}_{\alpha} \rangle) (\mathbf{r}_{\beta} - \langle \mathbf{r}_{\beta} \rangle) \rangle, \qquad (2)$$

where $\langle \cdots \rangle$ denotes the ensemble average, while $\mathbf{r}_{\alpha}(\mathbf{r}_{\beta})$ and $m_{\alpha}(m_{\beta})$ are the position vector and mass of particle $\alpha(\beta)$, respectively. Diagonalization of the matrix A,

$$E^{\mathsf{T}}\mathsf{A}\mathsf{E}=\boldsymbol{\lambda},\tag{3}$$

gives eigenvectors $E = (\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_{3N})$, where *N* stands for the number of particles, and the corresponding eigenvalues are $\lambda = (\lambda_1, \lambda_2, \dots, \lambda_{3N})$. The eigenvalue λ_n ($n = (1, \dots, 3N)$) is the mean-square fluctuation of the *n*th collective mode, which is related to its frequency as follows:

$$\omega_n^2 = \frac{k_B T}{\lambda_n},\tag{4}$$

where k_B and *T* stand for the Boltzmann constant and temperature, respectively.^{23,27,69}

PCA eigenvectors, \mathbf{e}_n , identifying the directions of particle's motion of the *n*th mode can be further projected along some predefined directions.⁷⁰ Such a mode-projection analysis can be used to elucidate the radial character of collective modes by introducing the time-independent quantity \mathcal{R} , defined as

$$\mathcal{R} = |\mathbf{e}_n \cdot \mathbf{d}_r| / |\mathbf{d}_r|,\tag{5}$$

where $|\cdots|$ stands for the absolute value and $\mathbf{d}_r = \langle \mathbf{r}_{\alpha} \rangle - \mathbf{r}_{cm}$ is the radial vector to the averaged particle position relative to the protein center of mass (CoM).

III. COMPUTATIONAL DETAILS

Employing OBMD allows us to impose acoustic waves with frequencies matching those of the protein's collective modes. The simulation box of dimensions $12.3 \times 6.2 \times 6.2$ nm³ is divided into two buffer regions, surrounding the ROI. The protein is placed at the center of the ROI and enclosed by water molecules (see Fig. 1). The CG Martini 3 model of ubiquitin is obtained after treating the initial atomistic structure (PDB entry 1UBQ⁷¹) with Martinize2, available at https://github.com/cmbi/dssp/releases/tag/2.3.0 (for illustration, see Fig. 2). The information on the classification of the secondary structure of the protein backbone from the structure is provided from the DSSP database available at https://github.com/cmbi/dssp/releases/tag/2.3.0. The protein's secondary and tertiary structures are maintained by additional harmonic bonds added between the backbone particles. The corresponding elastic bond constant is set to 550 kJ mol⁻¹ nm⁻², where the lower and upper elastic bond cutoffs are fixed at 0.5 and 0.9 nm, respectively. The geometry of the protein molecule is constrained using RATTLE.⁷² The equations of motion are integrated using the velocity Verlet algorithm⁷³ with an integration time step $\delta t = 0.02$ ps at a temperature of 300 K. All simulations are performed using the Martini v3.0.0 force field⁵⁶ and ESPResSo++ simulation package.⁷⁴ Non-bonded interactions are described using the Lennard-Jones 12-6 potential energy function with a cutoff value of 1.1 nm, while the long-range interactions of charged groups are treated by the



FIG. 2. A cartoon representation of the atomistic structure of ubiquitin on the left, accompanied by the CG Martini 3 model of ubiquitin on the right. To increase clarity, only backbone beads of the CG protein are depicted.

Coulomb energy function with a relative permeability $\varepsilon_r = 15$ and a cutoff distance of 1.1 nm. To sample a wide range of protein's conformational space in and out-of-equilibrium, several parallel simulations of length 500 ns are conducted, for which only the initial position of the protein is varied. The dynamic behavior of the biomolecule is analyzed over the entire trajectory and averaged over all trajectories sampled in equilibrium simulations or at a given frequency of the imposed acoustic wave. In addition, to prevent the protein from diffusing through the open ends of the simulation box or over the edges where the periodic boundary conditions are implemented, an additional harmonic constraint ($k = 3.5 \text{ kg s}^{-2}$) is added to hold the protein CoM close to the center of the ROI. As we are only interested in internal vibrational dynamics of ubiquitin, implementation of the harmonic constraint is not expected to affect the results.

IV. RESULTS AND DISCUSSION

We investigate the internal dynamics of the protein by subjecting it to acoustic waves with frequencies corresponding to those of its vibrational collective modes. To obtain collective modes, we perform OBMD simulations in equilibrium (where from both sides only a constant normal load is imposed on the system) and apply PCA. Given that the number of PCA modes scales with the number of beads taken into account and for the fact that we are interested in as few as possible representative modes, we perform PCA on further reduced coordinates representing CoM of six selected subunits. Accordingly, we obtain 18 modes. Figure 3 shows the lowest (left) and the highest (right) PCA modes of ubiquitin in a six-bead representation. In addition, localized motions in a few segments can be observed in Fig. 3 (right). As the low-frequency collective motions of the protein are usually related to its function,¹³⁻¹⁷ we focus on excitation of only the lowest collective modes of ubiquitin. To this end, we excite acoustic waves with frequencies up to 0.080 THz (which covers the frequency range of the four lowest vibrational modes). Parameters p and Δp are fixed at 1 and 100 bars, respectively.

The average acoustic energy density (\overline{w}) introduced into the system is estimated using $\overline{W} = j/c$, where *c* and *j* stand for the speed of sound and energy density of the flux, respectively. The latter is expressed as $j = 1/2(\Delta p)^2/\rho c$. Using the expression for \overline{w} for two different pressure amplitudes, i.e., 100 and 200 bars, the value of

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FIG. 3. Representations of displacement vectors of six subunits (spherical beads) in the lowest (left) and the highest (right) PCA mode of vibration in ubiquitin with frequencies of 0.009 and 0.351 THz, respectively. The color scheme is consistent with the one used in Fig. 2.

 \overline{w} is estimated to be ~17000 and 70500 kg m⁻¹ s⁻², respectively. Multiplying \overline{w} by the volume of interest, the kinetic energies at $\Delta p = 100$ and 200 bars are ~10⁻²² and ~10⁻²¹ kg m² s⁻², respectively. The acoustic energy pumped into the system turns out to be relatively small compared to the thermal energy stored in the same volume, which is ~10⁻¹⁹ kg m² s⁻². Nevertheless, the acoustic wave propagating in one direction clearly introduces anisotropy of particle velocity distribution by enhancing the longitudinal component, as shown in Fig. 4.

To infer the protein susceptibility to induced mechanical pressure waves, we compute displacements of the CoM of selected subunits from their average positions: $\rho_{\alpha}(t) = r_{\alpha}(t) - \langle r_{\alpha} \rangle$, where $r_{\alpha} = |\mathbf{r}_{\alpha}|$ is the position of the CoM of subunit $\alpha = (1, ..., 6)$ and $\langle \cdots \rangle$ denotes the ensemble average. In addition, using $R_g(t) = \sqrt{1/M\sum_{\alpha} m_{\alpha}(\mathbf{r}_{\alpha}(t) - \mathbf{r}_{cm}(t))}$, we calculate the radius of gyration, which provides information about the compactness of the protein structure. Here, $M = \sum_{\alpha} m_{\alpha}$ and \mathbf{r}_{cm} is the position vector of the protein's CoM.⁷⁵ These properties are used to compute generalized velocities $V_g(t) = dR_g(t)/dt \approx \Delta R_g/\Delta t$ and $V_{\alpha}(t)$



FIG. 4. Comparison of individual components W_k along the coordinate axes of the perturbed system with the kinetic energy of the reference system. The acoustic wave propagates along the *x*-axis. Kinetic energy is evaluated in the selected volume of interest, which is located in the ROI. The error bars represent the associated standard error of the mean.

 $= d\rho_{\alpha}(t)/dt \approx \Delta \rho_{\alpha}/\Delta t$, which we determine from molecular dynamics trajectory at time step $\Delta t = 5$ ps. They are further used to calculate the Fourier transforms of the autocorrelation functions providing the corresponding vibrational densities of states (vDOS),^{76,77}

$$g_{R_g/r_\alpha}(\nu) = \frac{1}{2\pi} \int \int V_{g/\alpha}(\tau+t) V_{g/\alpha}(\tau) d\tau \, e^{i2\pi\nu t} \, dt. \tag{6}$$

We focus on the difference between the computed vDOS obtained under the influence of acoustic waves and the reference $g^0_{R_g/r_\alpha}(v)$ determined by equilibrium simulations,

$$G_{R_g/r_\alpha}(\nu) = g_{R_g/r_\alpha}(\nu) - g_{R_g/r_\alpha}^0(\nu).$$
⁽⁷⁾

Results for different frequencies of acoustic waves are shown in the top and middle panels of Figs. 5 and 6. Inspecting the $G_{r_{\alpha}}(\nu)$ spectrum for different beads representing protein subunits, we observe a pronounced peak at the frequency corresponding to the one used to excite the biomolecule (see the top panels of Figs. 5 and 6, where the black dotted line indicates the acoustic wave frequency to which the protein is subjected). The top panel of Fig. 5 depicts the results obtained after subjecting the protein to the acoustic wave with a frequency of 0.048 THz. As can be seen, the subunits respond distinctly at the frequency of 0.048 THz. Observed response means that exposure of the protein to acoustic wave causes the



FIG. 5. Top and middle panels depict the computed spectra of general velocities, while the bottom panel shows the radial displacement projected along the low-frequency vibrational modes when an acoustic wave with a frequency of 0.048 THz is used to excite the biomolecule. The color scheme used in the top and bottom panels corresponds to the one used in Fig. 2. The colored areas and the error bars reoresent the standard error of the mean.



FIG. 6. Same as in Fig. 5 but for the frequency of 0.062 THz.

subunits to make larger deviations from their average positions. However, not all subunits respond to the same extent [for instance, compare $G_{r_a}(v)$ of different subunits in the top panel of Fig. 5]. This observation might be explained by arguing that the protein is expected to respond strongly to the acoustic wave excitation of a certain frequency if the direction of the vibrational mode is aligned with the propagating direction of the acoustic wave. Conversely, if the collective mode is oriented perpendicular to the direction of the acoustic wave propagation, no distinct response is expected. Nevertheless, vibrational modes whose displacements are in arbitrary directions contribute to response, but their contribution depends on the deviation from the direction of the wave propagation. Similarly, the highest peak (or one of the highest peaks) in $G_{R_a}(v)$ is observed at the excitation frequency (see the middle panels of Figs. 5 and 6). This further substantiates the sensitivity of the protein structure to the applied mechanical stress. In addition, the indicated changes in the compactness of the protein structure could also be due to the breathing-like dynamic character of the collective mode. For comparison, we add the results for the case in which the frequency of acoustic wave differs significantly from any of the mode frequencies of the protein in Fig. 7, demonstrating only a modest response in $G_{R_g/r_a}(v)$, which should serve as a "good negative control."

The computed \mathcal{R} values, defined in Eq. (5), are shown as the function of the collective mode frequency in the bottom panels of Figs. 5 and 6. In general, by comparing the \mathcal{R} values at the frequencies used to perturb the protein (depicted by the purple crosses) with the \mathcal{R} values of the protein in equilibrium (shown by the orange crosses), we find that the former increase. This suggests that the acoustic wave perturbation causes dynamic motion of



FIG. 7. Negative control showing a modest response to the acoustic wave with the frequency of 0.076 THz, which does not correspond to the protein's mode of vibration. The colored areas and the error bars represent the standard error of the mean.

the collective mode to become more breathing-like. As can also be observed, frequencies at which the breathing-like character of collective modes is present are correlated with frequencies at which peaks in $G_{R_g}(v)$ occur. Therefore, part of the change in the compactness of the protein structure could be attributed to the breathing-like dynamic motion. Bead representations of vibrational modes with the frequency corresponding to the frequency used for excitation are shown in the insets of the bottom panels of Figs. 5 and 6. As can be seen, displacement vectors in modes matching the excitation frequencies of 0.048 and 0.062 THz have a pronounced radial character, which is in agreement with higher \mathcal{R} values of collective modes depicted in Figs. 5 and 6 (see the purple crosses next to the black dotted lines).

In experimental setups applying electromagnetic THz waves, the most common parameters varied are the frequency of radiation, intensity, and duration of exposure to the terahertz radiation.^{50,78} For the implemented acoustic waves, we can vary all these parameters: the frequency (i.e., input parameter ν), the amplitude of the imposed wave (i.e., input parameter Δp), and the time of exposure (i.e., simulation length). By varying the frequency, we show that the protein is highly susceptible to induced mechanical stress, exhibiting larger deviations from average positions and increased breathing-like dynamic motion of modes. As we anticipate that the higher amplitude (i.e., Δp) would cause even larger displacements, we additionally perform OBMD simulations in the similar manner as before, except that we impose only acoustic waves with a frequency of 0.062 THz and set Δp to 200 bars.



FIG. 8. Comparison of vDOS derived from the displacements of six subunits from their average positions at amplitudes (Δp) 100 (solid line) and 200 (dashed line) bars, where a frequency of 0.062 THz is used to perturb the biomolecule. The color scheme follows the one used in Fig. 2. The standard error of the mean is ~0.01, reaching 0.03 at the peak.



FIG. 9. Comparison of vDOS derived from the radius of gyration when acoustic waves with a frequency of 0.062 THz at two different amplitudes are used to excite the biomolecule. The colored areas represent the standard error of the mean.

Repeating the analysis, we obtain the $G_{r_a}(\nu)$ spectra and compare them with those determined at the amplitude of 100 bars. The results are shown in Fig. 8. Taking these into consideration, even more pronounced peaks are observed at frequencies matching the excitation ones (indicated using the black dotted lines). This suggests that the beads describing protein subunits increasingly deviate from their average positions when subjected to mechanical pressure waves of higher amplitudes. Note that not all objects respond in the same manner due to the deviation of displacement vectors from the direction in which the imposed acoustic wave is propagated. Nevertheless, at higher amplitudes of the pressure wave, the protein is even more susceptible to mechanical stress, as supported by a higher peak in $G_{R_g}(\nu)$ (see Fig. 9).

V. CONCLUSIONS

In the present study, we show that the internal motion of a protein can be manipulated by mechanical pressure waves of different frequencies. To this end, a harmonically restrained coarse-grained protein model is immersed in water and subjected to acoustic waves whose frequencies are chosen around the frequencies of the protein's quasi-harmonic collective modes. Such a protein model is biased by its elastic character, which affects the resonant behavior of the protein compared to a real system dominated by highly anharmonic vibrations at low frequencies. Nevertheless, it allows straightforward application of the PCA to determine the collective motions of the protein. By analyzing the autocorrelation functions of the generalized velocities derived from displacements of beads that describe the protein subunits and from the radius of gyration, we show that the protein is susceptible to mechanical stress induced by acoustic waves. The response of the protein is more pronounced at frequencies corresponding to the acoustic wave. The assumption that the pressure wave with a larger amplitude would cause even larger deviations and enhanced changes in the protein compactness is confirmed and demonstrated by our simulations. Moreover, the dynamic response of the protein also depends on the correspondence between the direction of the corresponding collective mode and the propagation direction of the acoustic wave at a given frequency.

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AUTHOR DECLARATIONS

Conflict of Interest

The authors have no conflicts to disclose.

Author Contributions

Petra Papež: Formal analysis (lead); Methodology (equal); Validation (lead); Visualization (lead); Writing – original draft (lead). **Franci Merzel**: Conceptualization (equal); Formal analysis (equal); Methodology (equal); Validation (equal); Writing – original draft (supporting); Writing – review & editing (equal). **Matej Praprotnik**: Conceptualization (lead); Formal analysis (equal); Funding acquisition (lead); Methodology (equal); Supervision (equal); Validation (equal); Writing – original draft (supporting); Writing – review & editing (equal).

DATA AVAILABILITY

The data that support the findings of this study are available from the corresponding author upon reasonable request.

REFERENCES

- ¹C. B. Anfinsen, Science 181, 223 (1973).
- ²A. R. Fersht, Nat. Rev. Mol. Cell Biol. **9**, 650 (2008).
- ³A. J. Rader, C. Chennubhotla, L.-W. Yang, and I. Bahar, in *Normal Mode Analy*sis: Theory and Applications to Biological and Chemical Systems, edited by Q. Cui and I. Bahar (Chapman and Hall/CRC, Boca Raton, 2006), pp. 41-64.
- ⁴K. Henzler-Wildman and D. Kern, Nature 450, 964 (2007).
- ⁵ P. C. Biggin, Phys. Life Rev. **10**, 27 (2013).
- ⁶P. K. Agarwal, Phys. Life Rev. 10, 35 (2013).
- ⁷M. Kurplus and J. A. McCammon, Annu. Rev. Biochem. 52, 263 (1983).
- ⁸R. G. Smock and L. M. Gierasch, <u>Science</u> **324**, 198 (2009).
- ⁹A. Ramanathan, A. Savol, V. Burger, C. S. Chennubhotla, and P. K. Agarwal, Acc. Chem. Res. 47, 149 (2014).
- ¹⁰J. A. McCammon, Rep. Prog. Phys. 47, 1 (1984).
- ¹¹ K. A. Henzler-Wildman, M. Lei, V. Thai, S. J. Kerns, M. Karplus, and D. Kern, Nature 450, 913 (2007).
- 12K. A. Henzler-Wildman, V. Thai, M. Lei, M. Ott, M. Wolf-Watz, T. Fenn, E. Pozharski, M. A. Wilson, G. A. Petsko, M. Karplus, C. G. Hübner, and D. Kern, Nature 450, 838 (2007).
- ¹³Y. Seno and N. Gō, J. Mol. Biol. 216, 111 (1990).
- ¹⁴J. Ma and M. Karplus, Proc. Natl. Acad. Sci. U. S. A. 95, 8502 (1998).
- ¹⁵A. Thomas, K. Hinsen, M. J. Field, and D. Perahia, Proteins 34, 96 (1999).
- ¹⁶F. Tama and Y.-H. Sanejouand, Protein Eng., Des. Sel. 14, 1 (2001).
- ¹⁷Y.-H. Sanejouand, in Normal Mode Analysis: Theory and Applications to Biological and Chemical Systems, edited by Q. Cui and I. Bahar (Chapman and Hall/CRC, Boca Raton, 2006), pp. 91-109.
- ¹⁸H. J. Berendsen and S. Hayward, Curr. Opin. Struct. Biol. 10, 165 (2000).
- ¹⁹K. Hinsen, in Normal Mode Analysis: Theory and Applications to Biological and Chemical Systems, edited by Q. Cui and I. Bahar (Chapman and Hall/CRC, Boca Raton, 2006), pp. 1-16.
- ²⁰L. Skjaerven, S. M. Hollup, and N. Reuter, J. Mol. Struct.: THEOCHEM 898, 42 (2009).
- ²¹B. R. Brooks, D. Janežič, and M. Karplus, J. Comput. Chem. 16, 1522 (1995).
- ²²A. Amadei, A. B. M. Linssen, and H. J. C. Berendsen, Proteins 17, 412 (1993).
- 23 S. Hayward, A. Kitao, and N. Gō, Protein Sci. 3, 936 (1994).
- ²⁴S. Hayward, A. Kitao, and N. Gō, Proteins 23, 177 (1995).
- ²⁵S. Hayward and N. Go, Annu. Rev. Phys. Chem. 46, 223 (1995).
- ²⁶A. Amadei, B. de Groot, M.-A. Ceruso, M. Paci, A. Di Nola, and H. Berendsen, Proteins 35, 283 (1999).
- ²⁷ M. A. Balsera, W. Wriggers, Y. Oono, and K. Schulten, J. Phys. Chem. 100, 2567 (1996).
- ²⁸Y. He, J. Y. Chen, J. R. Knab, W. Zheng, and A. G. Markelz, Biophys. J. 100, 1058 (2011).
- ²⁹F. Bonnier and H. J. Byrne, Analyst **137**, 322 (2012).
- ³⁰A. Carpinteri, G. Lacidogna, G. Piana, and A. Bassani, J. Mol. Struct. 1139, 222 (2017).
- ³¹S. Cusack, J. Smith, J. Finney, B. Tidor, and M. Karplus, J. Mol. Biol. 202, 903 (1988).
- ³²W. Doster, S. Cusack, and W. Petry, Nature **337**, 754 (1989).
- ³³M. Diehl, W. Doster, W. Petry, and H. Schober, Biophys. J. **73**, 2726 (1997).
- ³⁴L. Hong, N. Jain, X. Cheng, A. Bernal, M. Tyagi, and J. C. Smith, Sci. Adv. 2, e1600886 (2016).
- ³⁵A. Ditta, H. Nawaz, T. Mahmood, M. I. Majeed, M. Tahir, N. Rashid, M. Muddassar, A. A. Al-Saadi, and H. J. Byrne, Spectrochim. Acta, Part A 221, 117173 (2019).
- ³⁶A. Markelz, S. Whitmire, J. Hillebrecht, and R. Birge, Phys. Med. Biol. 47, 3797 (2002).
- ³⁷M. Walther, P. Plochocka, B. Fischer, H. Helm, and P. Uhd Jepsen, Biopolymers 67, 310 (2002).
- ³⁸S. E. Whitmire, D. Wolpert, A. G. Markelz, J. R. Hillebrecht, J. Galan, and R. R. Birge, Biophys. J. 85, 1269 (2003).

- ³⁹K. G. Brown, S. C. Erfurth, E. W. Small, and W. L. Peticolas, Proc. Natl. Acad. Sci. U. S. A. 69, 1467 (1972).
- 40 L. Genzel, F. Keilmann, T. P. Martin, G. Wintreling, Y. Yacoby, H. Fröhlich, and M. W. Makinen, Biopolymers 15, 219 (1976).
- ⁴¹P. C. Painter, L. E. Mosher, and C. Rhoads, Biopolymers 21, 1469 (1982).
- ⁴²S. E. May Colaianni, J. Aubard, S. Høime Hansen, and O. Faurskov Nielsen, Vib. Spectrosc. 9, 111 (1995).
- ⁴³C. Crupi, G. D'Angelo, U. Wanderlingh, and C. Vasi, Philos. Mag. 91, 1956 (2011).
- ⁴⁴C. V. Berney, V. Renugopalakrishnan, and R. S. Bhatnagar, Biophys. J. 52, 343 (1987).
- ⁴⁵S. Cusack and W. Doster, Biophys. J. 58, 243 (1990).
- ⁴⁶H. D. Middendorf, R. L. Hayward, S. F. Parker, J. Bradshaw, and A. Miller, Biophys. J. 69, 660 (1995).
- ⁴⁷H. H. Mantsch and D. Naumann, J. Mol. Struct. 964, 1 (2010).
- ⁴⁸L. Wei, L. Yu, H. Jiaoqi, H. Guorong, Z. Yang, and F. Weiling, Front. Lab. Med. 2, 127 (2018).
- ⁴⁹B. S. Kalanoor, M. Ronen, Z. Oren, D. Gerber, and Y. R. Tischler, ACS Omega 2, 1232 (2017).
- ⁵⁰S. Yamazaki, M. Harata, T. Idehara, K. Konagaya, G. Yokoyama, H. Hoshina, and Y. Ogawa, Sci. Rep. 8, 9990 (2018).
- ⁵¹S. Yamazaki, M. Harata, Y. Ueno, M. Tsubouchi, K. Konagaya, Y. Ogawa, G. Isoyama, C. Otani, and H. Hoshina, Sci. Rep. 10, 9008 (2020).
- ⁵²H. Hoshina, S. Yamazaki, M. Tsubouchi, and M. Harata, J. Phys. Photonics 3, 034015 (2021).
- ⁵³M. Tsubouchi, H. Hoshina, M. Nagai, and G. Isoyama, Sci. Rep. 10, 18537 (2020).
- ⁵⁴V. M. Govorun, V. E. Tretiakov, N. N. Tulyakov, V. B. Fleurov, A. I. Demin, A. Yu. Volkov, V. A. Batanov, and A. B. Kapitanov, Int. J. Infrared Millimeter Waves 12, 1469 (1991).
- ⁵⁵P. R. Vlachas, J. Zavadlav, M. Praprotnik, and P. Koumoutsakos, J. Chem. Theory Comput. 18, 538 (2022).
- ⁵⁶P. C. T. Souza, R. Alessandri, J. Barnoud, S. Thallmair, I. Faustino, F. Grünewald, I. Patmanidis, H. Abdizadeh, B. M. H. Bruininks, T. A. Wassenaar, P. C. Kroon, J. Melcr, V. Nieto, V. Corradi, H. M. Khan, J. Domański, M. Javanainen, H. Martinez-Seara, N. Reuter, R. B. Best, I. Vattulainen, L. Monticelli, X. Periole, D. P. Tieleman, A. H. de Vries, and S. J. Marrink, Nat. Methods 18, 382 (2021).
- ⁵⁷E. G. Flekkøy, R. Delgado-Buscalioni, and P. V. Coveney, Phys. Rev. E 72, 026703 (2005).
- ⁵⁸R. Delgado-Buscalioni, in Numerical Analysis of Multiscale Computations, Lecture Notes in Computational Science and Engineering, edited by B. Engquist, O. Runborg, and Y.-H. R. Tsai (Springer, Berlin, Heidelberg, 2012), pp. 145-166.
- ⁵⁹R. Delgado-Buscalioni, J. Sablić, and M. Praprotnik, Eur. Phys. J.: Spec. Top. 224, 2331 (2015).
- ⁶⁰ J. Sablić, M. Praprotnik, and R. Delgado-Buscalioni, Soft Matter **12**, 2416 (2016). ⁶¹G. De Fabritiis, R. Delgado-Buscalioni, and P. V. Coveney, Phys. Rev. Lett. **97**, 134501 (2006).
- ⁶²P. Papež and M. Praprotnik, J. Chem. Theory Comput. 18, 1227 (2022).
- ⁶³L. Delle Site and M. Praprotnik, Phys. Rep. **693**, 1 (2017).
- ⁶⁴ P. Español and P. Warren, Europhys. Lett. 30, 191 (1995).
- ⁶⁵R. D. Groot and P. B. Warren, J. Chem. Phys. 107, 4423 (1997).
- 66 T. Soddemann, B. Dünweg, and K. Kremer, Phys. Rev. E 68, 046702 (2003).
- ⁶⁷C. C. David and D. J. Jacobs, in *Protein Dynamics: Methods and Protocols*, edited by D. R. Livesay (Humana Press, Totowa, NJ, 2014), pp. 193-226.
- ⁶⁸S. Hayward and B. L. de Groot, in *Molecular Modeling of Proteins*, edited by A. Kukol (Humana Press, Totowa, NJ, 2008), pp. 89-106.
- ⁶⁹J. C. Smith, F. Merzel, A.-N. Bondar, A. Tournier, and S. Fischer, Philos. Trans. R. Soc. London, Ser. B 359, 1181 (2004).
- ⁷⁰ F. Merzel, F. Fontaine-Vive, M. R. Johnson, and G. J. Kearley, Phys. Rev. E 76, 031917 (2007).
- ⁷¹S. Vijay-Kumar, C. E. Bugg, and W. J. Cook, J. Mol. Biol. **194**, 531 (1987).
- ⁷²H. C. Andersen, J. Comput. Phys. 52, 24 (1983).
- ⁷³M. P. Allen and D. J. Tildesley, *Computer Simulation of Liquids* (Clarendon Press, 1987).

J. Chem. Phys. 159, 135101 (2023); doi: 10.1063/5.0163801 © Author(s) 2023

159, 135101-7

⁷⁴J. D. Halverson, T. Brandes, O. Lenz, A. Arnold, S. Bevc, V. Starchenko, K. Kremer, T. Stuehn, and D. Reith, Comput. Phys. Commun. 184, 1129 (2013).
⁷⁵M. Y. Lobanov, N. S. Bogatyreva, and O. V. Galzitskaya, Mol. Biol. 42, 623 (2008).

⁷⁶M. Ceriotti, G. Bussi, and M. Parrinello, J. Chem. Theory Comput. **6**, 1170 (2010).

77 D. Donadio and G. Galli, Nano Lett. 10, 847 (2010).

⁷⁸L. Sun, L. Zhao, and R.-Y. Peng, Mil. Med. Res. 8, 28 (2021).