

Introduction

- Proteins have intrinsic global dynamics encoded by their structure and important for their function [1]
- Normal mode analysis (NMA)** with coarse grained elastic network models (ENMs) is a good method for describing such dynamics [1]
- Traditional structural biology methods, including CryoEM with discrete maximum likelihood classification, provide preferred conformational states with differences between them that are in line with NMA
- This information can be harnessed to extract continuous dynamics and landscapes using new CryoEM image processing methods
- We demonstrate the beginnings of such a method, building upon our recent **Zernikes3D** method [2]

Theory

- Conformational changes between two structures can be described by deformation fields comprised of a basis set and corresponding coefficients
- Zernikes3D** [2] approximates a volume $V'(r)$ as the deformation of a reference volume $V(r)$ as a linear combination of the 3D Zernike polynomials $Z_{lmm}(r)$ at each voxel r with coefficients α_{lmm}
- These polynomials can also be calculated for atoms
- NMA eigenvectors can also be projected onto a deformation vector to approximate conformational change as a linear combination of mode vectors u_k with scaling coefficients g_k
- Hence, there's an equivalence between the two sets when considering structures with P atoms

$$\begin{pmatrix} U_1 \\ U_2 \\ \vdots \\ U_P \end{pmatrix} \mathbf{g} = (I_P \otimes A) \begin{pmatrix} Z_1 \\ Z_2 \\ \vdots \\ Z_P \end{pmatrix} \text{ or } (I_P \otimes \mathbf{g}^T) \begin{pmatrix} U_1^T \\ U_2^T \\ \vdots \\ U_P^T \end{pmatrix} = \begin{pmatrix} Z_1^T \\ Z_2^T \\ \vdots \\ Z_P^T \end{pmatrix} A^T$$

and we can solve it by least squares:

$$\mathbf{g} = (U^T U)^{-1} U^T (I_P \otimes A) Z$$

and

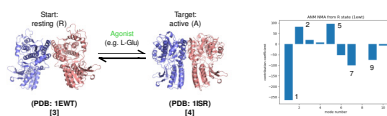
$$A^T = (Z^T Z)^{-1} Z^T (I_P \otimes \mathbf{g}^T) U^T$$

Results

1. NMA Projection Coefficients capture Transitions

- The simplest approximation of the transition between 2 structures of P atoms comes from subtraction of atom positions and is called the deformation vector
$$\Delta \mathbf{q} = [\Delta x_1 \ y_1 \ z_1 \ \Delta x_2 \ \dots \ \dots \ \dots \ \Delta z_P]^T$$
- (Linear interpolation along this vector gives a morph)
- Normal mode vectors have the same form with x, y, and z movement components for each atom
- A dot product between normal modes (NMs) and the deformation vector yields projection coefficients g_k

Example 1: metabotropic glutamate receptor (mGluR)

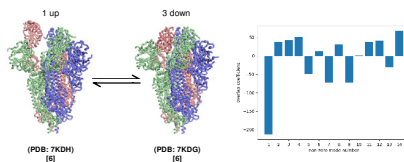


- Activation steps for Venus fly trap (VFT) dimer upon agonist binding [3, 4] are captured by global modes [5]:

- Mode 1: inter-subunit rotation
- Mode 2: inter-subunit closure
- Mode 5: closing of clamshell B
- Mode 7: closing of clamshell A
- Mode 9: symmetric clamshell tilting

- With 10 NMs, RMSD to target drops from 11.2 Å → 3.4 Å
- With 100 NMs, RMSD to target drops from 11.2 Å → 1.2 Å

Example 2: SARS-CoV-2 Spike



- Receptor-binding domains (RBDs) alternate between an "up" state to engage host cells and a "down" state that evades the immune system, and can be impacted by mutations including D614G [6-8]

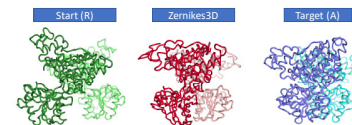
- A single mode captures the RBD coming down
- RMSD to target drops from 5.9 Å → 3.0 Å

2. Conversion of Coefficients Helps Zernikes3D

- Conversions can be applied in either direction
- These are illustrated with the same two examples

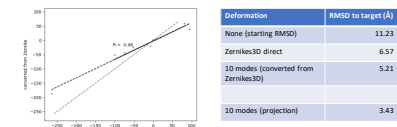
Example 1: metabotropic glutamate receptor (mGluR)

- The Zernikes3D atoms and volumes programs can capture the VFT dimer transition reasonably well



- Converting to NMA removes unphysical deformations by not having modes that include them

- It recovers the same modes as the projection with similar weights but smaller amplitudes
- RMSD improves relative to Zernikes3D alone although still worse than NMA projection

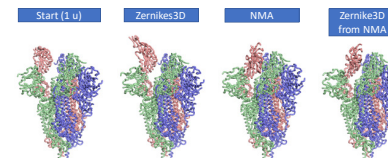


- This could be useful for selecting modes to save efficiency in hybrid electron microscopy normal mode analysis (HEMNMA) continuous heterogeneity analysis and for flexible fitting and other map-based simulations

Example 2: SARS-CoV-2 Spike

- Zernikes3D has difficulty with RBD up/down motion, even using higher order polynomials

- Converting from NMA to Zernikes3D really helps



- The improved 3D Zernike polynomials could be used as priors for continuous heterogeneity analysis and improving reconstructions of less populated states.

Methods

- Atomic structures were parsed and handled, and normal modes, deformation vectors, and projections were calculated with ProDy 2
- Zernikes3D command line programs in Xmipp were used directly, including atoms program for mGluR VFT and volumes program for the SARS-CoV-2 Spike
- Volumes of the Spike were simulated by converting atomic structures with Xmipp programs using Scipion
- Original volumes from the EMDDB were also tried with similar results
- Prototype code was developed in Python
- Key steps include:
 - Parsing of NMA vectors and Zernike coefficients
 - Calculation of NMA coefficients and 3D Zernike polynomials
 - Least squares conversion
 - Application of deformations
 - Analysis and output of new coefficients and deformed structures

Conclusions

- Normal modes and 3D Zernike polynomials can both work well for describing conformational changes of biological macromolecules
- A conversion between them using least squares enables their use in more circumstances and will facilitate further methods development
- This conversion is particularly helpful for the SARS-CoV-2 Spike, which causes difficulties for the Zernikes3D method

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