

Iterative Elastic 3D-to-2D Alignment Method Using Normal Modes for Studying Structural Dynamics of Large Macromolecular Complexes

Qiyu Jin,^{1,6} Carlos Oscar S. Sorzano,² José Miguel de la Rosa-Trevín,² José Román Bilbao-Castro,³ Rafael Núñez-Ramírez,^{4,7} Oscar Llorca,⁴ Florence Tama,⁵ and Slavica Jonić^{1,*}

¹IMPMC, Sorbonne Universités–CNRS UMR 7590, UPMC Paris 6, MNHN, IRD UMR 206, 4 Place Jussieu, 75005 Paris, France

²Biocomputing Unit, Centro Nacional de Biotecnología–CSIC, Madrid 28049, Spain

³Departamento de Informática, Universidad de Almería, Almería 04120, Spain

⁴Centro de Investigaciones Biológicas, CSIC, Madrid 28040, Spain

⁵RIKEN, Advanced Institute for Computational Sciences, Kobe, Hyogo 650-0047, Japan

⁶Present address: Institute of Image Processing and Pattern Recognition, Shanghai Jiao Tong University, Shanghai 200240, China

⁷Present address: Instituto de Estructura de la Materia, CSIC, Madrid 28006, Spain

*Correspondence: slavica.jonic@impmc.upmc.fr

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SUMMARY

This article presents a method to study large-scale conformational changes by combining electron microscopy (EM) single-particle image analysis and normal mode analysis (NMA). It is referred to as HEMNMA, which stands for hybrid electron microscopy normal mode analysis. NMA of a reference structure (atomic-resolution structure or EM volume) is used to predict possible motions that are then confronted with EM images within an automatic iterative elastic 3D-to-2D alignment procedure to identify actual motions in the imaged samples. HEMNMA can be used to extensively analyze the conformational changes and may be used in combination with classic discrete procedures. The identified conformations allow modeling of deformation pathways compatible with the experimental data. HEMNMA was tested with synthetic and experimental data sets of *E. coli* 70S ribosome, DNA polymerase Pol α and B subunit complex of the eukaryotic primosome, and tomato bushy stunt virus.

INTRODUCTION

Normal mode analysis (NMA) of structures relies on modeling complex dynamics by a linear combination of harmonic oscillations around an equilibrium structural conformation. NMA was extensively used in electron microscopy (EM), for predicting functional motions from non-atomic-resolution EM structures (Chacón et al., 2003; Ming et al., 2002; Tama et al., 2002) and for flexible fitting (Suhre et al., 2006; Tama et al., 2004a, 2004b) of atomic-resolution structures (e.g., from X-ray crystallography) into EM structures (EM density volumes). Normal modes of EM density volumes were shown to provide a good approximation of atomic-resolution normal modes in the low-frequency range that generally contains the modes reflecting

experimentally observed large-scale conformational changes (Tama et al., 2002). They are thus useful for predicting conformational dynamics of complexes whose structure at atomic resolution is unavailable, but an intermediate-resolution structure can be obtained by EM (Chacón et al., 2003; Ming et al., 2002; Tama et al., 2002).

Single-particle analysis (SPA) EM techniques have been traditionally used in determining structures of large macromolecular complexes (diameter, 10–30 nm) from samples prepared biochemically to ideally contain all complexes in the same steady conformation. The high homogeneity of conformations is required for a high resolution of the structure reconstructed from images. As the homogeneous population is sometimes difficult to obtain, even with the most improved biochemical procedures, image-processing methods have been developed to separate images into a few classes and determine the ones corresponding to the steady conformations that are being studied (Elad et al., 2008; Grob et al., 2006; Penczek et al., 2006; Scheres et al., 2007; Simonetti et al., 2008). The goal of these standard approaches is thus to identify the most homogeneous classes of images in terms of conformations and compute the average structure from the classes. More accurate classification methods combined with the biochemical sample preparation to reduce the number of possible conformations should result in a higher homogeneity of classes and a higher resolution of computed structures. As the expected number of conformations is small in such cases, the standard methods are generally used with a small number of classes defined by the user initially. The standard methods thus allow the study of only a few discrete conformations of the complex, and new developments are needed for studying a large range of conformations with many intermediate conformations. This problem arises when studying a full dynamics of the complex that can freely change the conformation or when the biochemical stabilization of the complex in a few steady conformations is impossible. In such cases, the standard methods provide the dynamics information biased by the mentioned initial discretization into a small number of classes (the reconstructed structures are not necessarily discrete samples of the conformational change, but they may result from mixing different conformations into the selected number of classes).

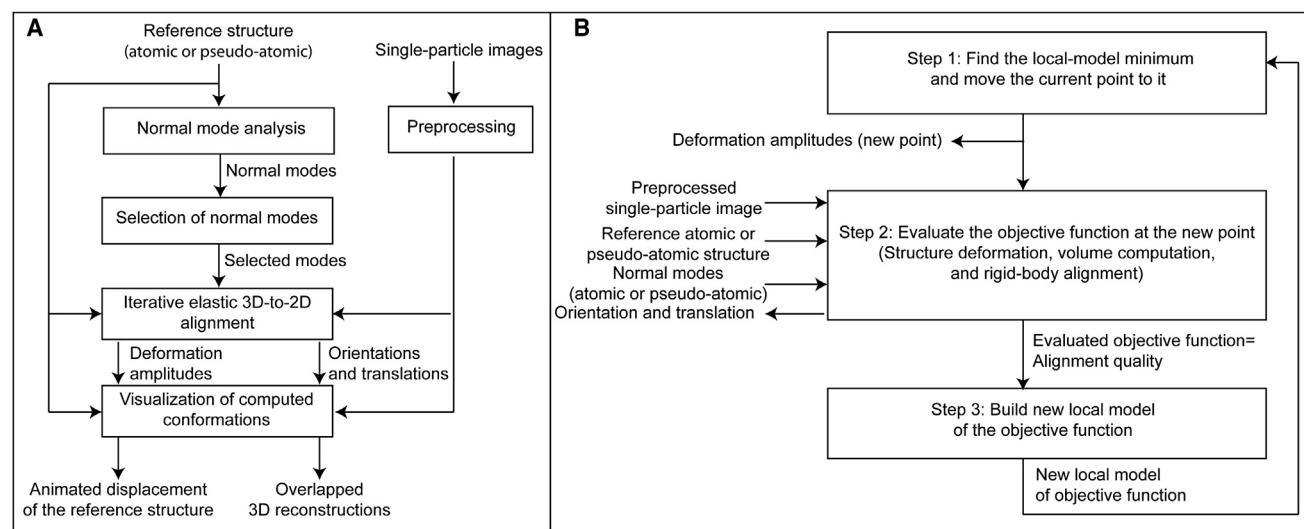


Figure 1. HEMNMA

Flowchart of the method for studying macromolecular dynamics (A) using a 3D-to-2D normal-mode-based alignment of atomic or pseudo-atomic structures with single-particle images (B).

In this article, we propose a method that was specifically designed for studying a large range of conformations with many intermediate conformations. It allows the analysis of highly heterogeneous populations of complexes with slow and continuous conformational changes (gradual conformational changes where a discrete number of states cannot fully describe the continuum of conformations). Given a reference structure (an atomic-resolution structure or a reduced representation of an EM structure), a set of normal modes of the structure, and a set of images, projection-matching-based elastic alignment of each single-particle image with the reference structure provides an overall view of the conformational distribution. A potential of combining NMA with EM image analysis was demonstrated by Brink et al. (2004), who compared several different conformations observed in images with the conformational changes predicted by NMA, through a visual evaluation of their consistency. Here, we describe an automatic method to determine conformations from images using normal modes. The proposed method is based on a procedure for iterative elastic projection matching. This procedure allows the processing of a high number of images to determine all actual conformations and evaluate their pertinence, which is impossible to achieve with other EM methods or X-ray crystallography. The proposed method has the following four main features: (1) the conformational modeling is performed by displacing the reference structure along combinations of normal modes, and the amplitudes of the displacement are computed through the elastic image alignment; (2) each single-particle image may represent a unique conformation; (3) classification and 3D reconstruction are not mandatory; and (4) classic discrete EM methods may be used to compute the reference structure when it is unavailable at atomic resolution and/or to identify a subset of images to analyze. Thanks to these features, this method allows modeling of deformation pathways compatible with the experimental data and analyzing the conformational changes more extensively than

using the classic discrete EM methods. The proposed methodology is referred to as hybrid EM NMA (HEMNMA). It was tested here using several synthetic and experimental data sets: synthetic and EM images of *E. coli* 70S ribosome (70S), EM images of DNA polymerase Pol α and B subunit complex (Pol α -B) of the eukaryotic primosome, and EM images of tomato bushy stunt virus (TBSV).

RESULTS

HEMNMA Design

The workflow (Figure 1A) comprises the following four steps: (1) NMA of a reference structure at atomic or pseudoatomic resolution (see Experimental Procedures for the volume-to-pseudoatoms conversion and NMA methods used here); (2) selection of normal modes; (3) iterative elastic 3D-to-2D alignment of single-particle images with the reference structure; and (4) visualization of computed conformations. The preprocessing block in Figure 1A may include any preprocessing of single-particle images (e.g., correction of the effects of the electron-microscope contrast transfer function [CTF]).

Selection of Normal Modes

The computation of normal modes takes negligible time with respect to the analysis of a long series of images using normal modes (e.g., 10,000 single-particle images). To reduce image analysis time while focusing on the most biologically relevant conformational changes, normal modes can be selected on the basis of some a priori knowledge about the conformational change (e.g., two known motions of 70S, ratchet-like and L1-stalk motions, were used in experiments 1 and 2). If no a priori knowledge is available, lowest frequency modes with the collectivity degree above a threshold can be selected, as highly collective low-frequency modes have been shown to be relevant to functional conformational changes (Delarue and Dumas, 2004; Suhre et al., 2006; Tama and Brooks, 2006; Tama et al.,

2004a; Tama and Sanejouand, 2001; Wang et al., 2004). For instance, the overlap between experimentally observed conformational changes (difference between two atomic conformations of the same complex) and the modes computed for one of the conformations of the complex has showed that 1–3 low-frequency modes with the highest overlap are highly collective and usually capture 60% to 70% of the conformational change (Tama and Sanejouand, 2001; Tama et al., 2002). The most relevant modes are generally among the 20 lowest frequency modes for low-symmetry structures (Tama and Sanejouand, 2001; Tama et al., 2002). They move to higher frequencies (due to degeneracy) for symmetric structures but remain among the 100 lowest frequency modes, even for highly symmetric structures such as icosahedral viruses (Tama and Brooks, 2005). The number of lowest frequency modes to compute can thus generally be fixed to 100. The collectivity degree is computed to count the number of (pseudo-)atoms that are significantly affected by the mode (Bruschweiler, 1995). Given the number of (pseudo-)atoms N , the collectivity degree is normalized between $1/N$ and 1. The collectivity degree approaches 1 for maximally collective movements, whereas for localized motions (few atoms move), it approaches 0. The most collective modes capturing functional conformational changes can generally be selected by setting the collectivity threshold to 0.5 (experiment 3). Higher values of the threshold should be chosen to select a reduced number of modes for highly symmetric structures (experiment 4). For instance, in the case of “swelling” icosahedral-symmetry viruses, many modes have collectivities above 0.5 (experiment 4). However, a few of them are usually enough to describe functional conformational changes (e.g., the mode describing best the experimentally observed conformational changes is related to a radially symmetric expansion of the capsid) (Tama and Brooks, 2005) and can be selected using the collectivity threshold of 0.75 (experiment 4).

Iterative Elastic 3D-to-2D Alignment Method

The amplitudes of the reference-structure displacement (deformation) along normal modes and the orientation and the position of the deformed structure are refined until a projection of the deformed-reference density volume is found that is the most similar to the analyzed single-particle image. The method is based on a local-search Powell's UOBYQA optimization (Powell, 2002) that uses quadratic approximation for building a local model of the objective function. It consists of repeating the following three steps iteratively until the change in elastic parameters (deformation amplitudes) and rigid-body parameters (orientation and position of the deformed structure) between two successive iterations becomes insignificant or a specified number of iterations is reached (Figure 1B):

Step 1: Find the minimum of the current local model of the objective function and move the current point (current vector of deformation amplitudes) to this minimum.

Step 2: Evaluate the objective function at the new vector of deformation amplitudes (new point). The structure is first deformed with the new deformation amplitudes, and the modified atomic or pseudoatomic coordinates are converted into a volume. The computation of volumes from pseudoatomic coordinates is based on the volume representation by a sum of Gaussian functions (Nogales-Cadenas et al.,

2013). The computation of volumes from atomic coordinates uses electronic-form atomic factors (Peng et al., 1996). The volume is subsequently aligned (rigid-body) with the single-particle image using a rough alignment based on a discrete library of reference projections of the volume (Sorzano et al., 2004a) and a continuous refinement using a gradient-based approach (Jonić et al., 2005). The alignment quality determined by a measure of similarity between the single-particle image and the corresponding best matching volume projection is considered as the new evaluation of the objective function for the optimization.

Step 3: Reconstruct the local model of the objective function around the new point with the new evaluation. Go back to Step 1.

The source code of the method is available in the open-source software Xmipp (Sorzano et al., 2004b), starting from the version Xmipp 3.0.1. The current implementation takes, on average, around 2, 4, and 6 min to analyze an image of size 128×128 pixels using one, two, and three normal modes, respectively, on a core of a Dual Intel Xeon X5472 processor (3.00GHz). Note that the execution time was measured with one to three modes, but the program allows the use of any number of modes. An MPI-parallelized version of the method allows an efficient analysis of a large number of images (xmipp_mpi_nma_alignment whose tutorial is at http://xmipp.cnb.csic.es/twiki/bin/view/Xmipp/Nma_alignment_v3), which was tested on standard-size clusters and supercomputer centers.

Visualization of Computed Conformations

The computed conformations can be visualized by displacing the reference atomic or pseudoatomic structure along normal modes using the amplitudes of the displacement computed through the elastic image alignment (direct visualization). Also, the animated volumes can be obtained by converting displaced atoms or pseudoatoms into volumes, as in Step 2 of the elastic 3D-to-2D alignment method. Optionally, 3D reconstruction from images can be obtained by separating images into classes with similar conformations (similar values of the computed displacement amplitudes). However, for highly heterogeneous samples, the underlying dynamics can only be partially elucidated by analyzing these reconstructed structures due to a discretization of the conformational distribution (discrete number of classes) and averaging during the reconstruction. A direct visualization shall be a preferred choice in such cases, as we show with Pol α -B (experiment 3). For more details on the visualization methods, see the corresponding section of [Supplemental Experimental Procedures S1](#) available online.

Experiment 1: Synthetic Continuous-type Conformational Change

In this experiment, we show robustness of HEMNMA to noise and to the CTF effects remaining after the phase inversion that is usually used for CTF correction. A variety of conformations were sampled from a synthetic continuous-type conformational change determined by deforming an atomic-resolution *E. coli* 70S structure (Figure S1A) referred to as 3I1OP (without ligands, in the conformation in which the 30S subunit is nonrotated with respect to the 50S subunit) (Zhang et al., 2009). Synthetic images [size, 128^2 pixels; pixel size, $(3 \text{ \AA})^2$] were computed as 300

random projections of the atomic structure displaced with random displacement amplitudes along two of its normal modes while preserving a constant ratio between the two displacement amplitudes (for details on the random image generation with a linear relationship between the contributions of the two modes, see [Supplemental Experimental Procedures S2](#)). The used modes were mode 11 as it describes the ratchet-like motion (rotation of 30S with respect to 50S visible in [Movie S1](#)) and mode 12 as it describes the L1-stalk motion. The modes responsible for these two motions have already been characterized computationally ([Tama et al., 2003](#); [Wang et al., 2004](#)), and they were chosen here as the combined motion was already observed experimentally ([Fu et al., 2011](#); [Valle et al., 2003](#)). The overall characteristics of the modes used in this article are similar to those reported previously, although the mode numbers are not the same because of different structures and methods used for calculations.

The amount of noise before and after CTF was adjusted for each tested defocus value (0.5 μm , 1 μm , and 2 μm) so that the images were obtained with the signal-to-noise ratio (SNR) of 0.3 or 0.1. The different image sets (one without noise and six for different defocus and SNR values; [Figure S1D](#)) were analyzed using only the mode capturing ratchet-like motion to see whether the proposed method can recover both synthesized motions: ratchet-like and L1-stalk motions. The elastic alignment of images was performed with respect to the given atomic structure and the pseudoatomic structure obtained from a synthetic volume of the atomic structure ([Figure S1C](#)). [Movie S2](#) shows the animated pseudoatomic mode 13 (ratchet-like mode) used in image analysis. Details of the pseudoatomic structure and normal modes computations can be found in [Supplemental Experimental Procedures S2](#).

The image alignment with a pseudoatomic structure is more challenging than the alignment with an atomic structure, as pseudoatomic and atomic normal modes have different resolutions and the same movement often appears at a different frequency in the two sets of modes ([Tama et al., 2002](#)). Here, a more difficult alignment in the pseudoatomic case was expected also because different structures were used to generate images (atomic structure) and to analyze them (pseudoatomic structure), while the same structure was used to generate and analyze images in the atomic case. However, such a difficulty is expected in practice, as the resolution of experimental images is usually higher than the resolution of the reference structure when using EM structures. The exploration of only one movement instead of two that are present in images is also challenging but close to realistic as, in practice, images are expected to be analyzed using only a few normal modes (sometimes, only one as in experiment 2). Interestingly enough, although the used mode was selected on the basis of previously observed 70S motions instead of using the modes collectivity criterion, this mode is actually the most collective one for each of the used 70S structures in experiments 1 and 2.

The results in the atomic and pseudoatomic cases show that the accuracy of estimation of rigid-body and elastic parameters decreases as SNR decreases and the defocus increases. Although the maximum value (over seven image sets) of the relative deformation-amplitude error doubled in the pseudoatomic case compared to the atomic case (8.9% for pseudoa-

toms with respect to 4.2% for atoms; [Table S1A](#)), the accuracy is globally good, with the absolute angular and shift errors that are similar in the two cases (below 0.6° and 0.15 pixels for pseudoatoms; below 0.5° and 0.15 pixels for atoms; [Table S1B](#)). In the most difficult case (SNR = 0.1, defocus = 2 μm), the root-mean-square deviation between the structures with computed and ground-truth deformation amplitudes along the mode describing the ratchet-like motion is about 0.46 Å for pseudoatoms and 0.22 Å for atoms, which is far below the commonly achievable resolution in EM. [Table S1C](#) summarizes how the key parameters of the volume-to-pseudoatoms conversion and NMA methods and the resulting alignment error depend on the volume approximation error (for more details on the influence of these parameters, see [Supplemental Experimental Procedures S2](#)).

Experiment 2: Mixed EM Images of *E. coli* 70S Ribosome with and without EF-G

In this experiment, HEMNMA was used to analyze cryo-EM images of a mixture of *E. coli* 70S ribosome with and without Elongation Factor G (EF-G) bound. These images were used in previous EM studies where ratchet-like and L1-stalk motions were observed on the reconstructions corresponding to the ribosome with and without EF-G ([Elad et al., 2008](#); [Scheres et al., 2007](#)). Here, the images were analyzed with the ratchet-like mode and separated into the classes with and without EF-G using the computed deformation amplitudes along the mode. Indeed, our tests with synthetic data (generated using EM structures of three different 70S conformations) showed that this mode alone can be used to classify images of the ribosome at distinct states of the conformational change combining ratchet-like and L1-stalk motions and to reconstruct the structure at distinct states of the conformational change (for details of these tests, see [Supplemental Experimental Procedures S3](#); [Figure S2](#)).

The data set contained 10,000 cryo-EM images [size, 130² pixels; pixel size, (2.82 Å)²] and a result of a supervised image classification into two subsets, with or without EF-G (5,000 images per subset). Ribosomes without EF-G contain three tRNAs in the classical A/A, P/P, and E/E positions and are in the normal (unratcheted) conformation. Ribosomes with EF-G contain a single tRNA in the hybrid P/E position and are ratcheted. The unratcheted-conformation EM structure EMD-5262 ([Fu et al., 2011](#)) was used to compute the pseudoatomic structure and its normal modes. Details of the computations are given in [Supplemental Experimental Procedures, S3](#), as the same structure and normal modes were used there with synthetic images capturing ratchet-like and L1-stalk motions from authentic EM structures. Image analysis was conducted using the pseudoatomic structure and its mode 9 that describes ratchet-like motion ([Supplemental Experimental Procedures S3](#); [Movie S3](#)).

The histogram of the computed amplitudes along the pseudoatomic mode 9 (green, in [Figure 2A](#)) may be interpreted as a mixture of at least two Gaussian-type distributions, as two peaks can be noticed. [Figure 2A](#) also shows the histogram portions corresponding to the image separation by supervised classification (with EF-G, red; without EF-G, blue). The bins containing one third of the images with the smallest amplitudes (mean amplitude, -401) are indicated by red asterisks, while the bins

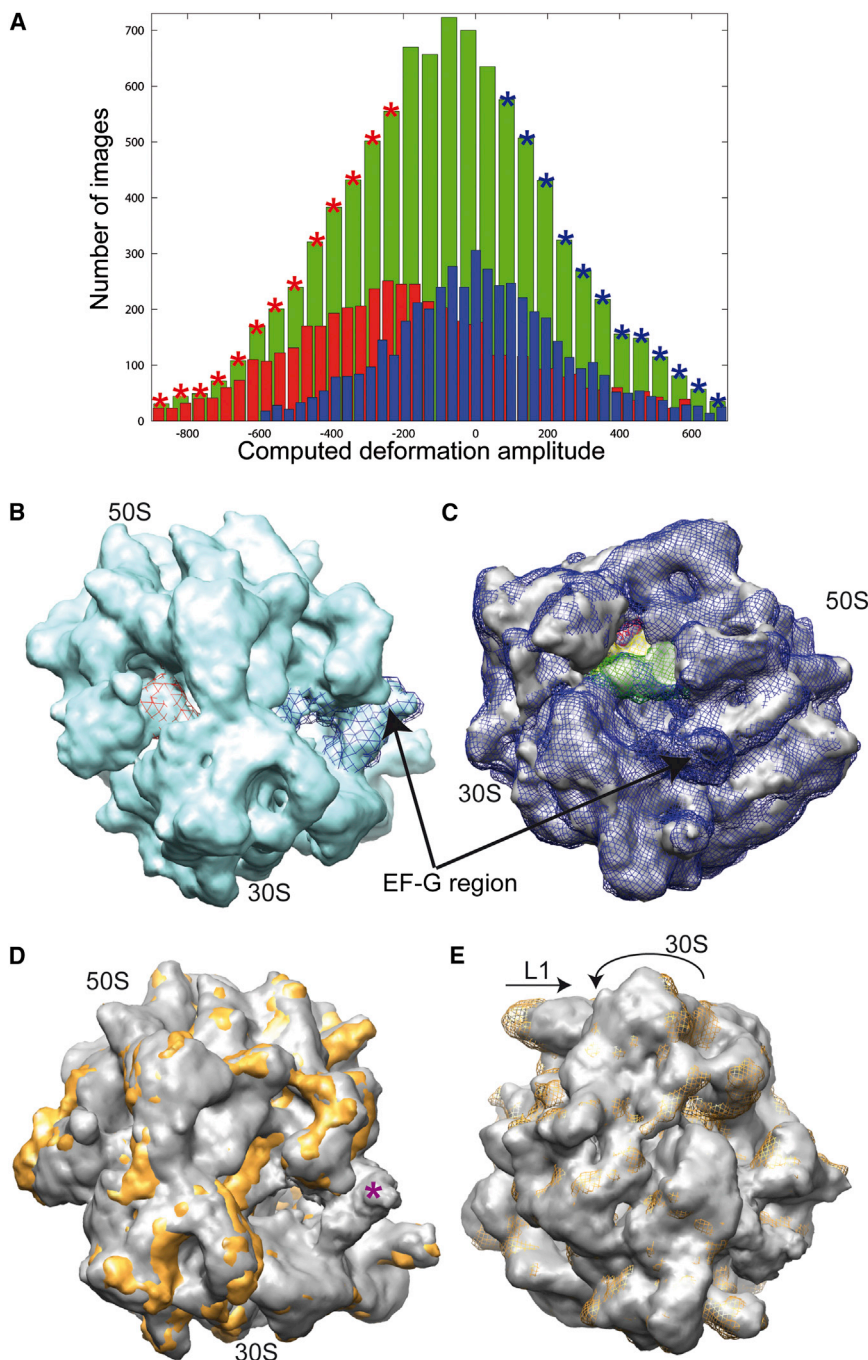


Figure 2. Experiment with EM Images of a Mixture of *E. coli* 70S Ribosome with and without EF-G Factor

(A) Histogram of computed deformation amplitudes (green) with the contribution of images classified by supervised classification as without EF-G (nonrotated conformation) and with EF-G (rotated conformation) indicated in blue and red, respectively. The bins corresponding to the images used for computing the two 3D reconstructions shown in (B)–(E) are indicated by two different colors of asterisks (red and blue).

(B) Structure reconstructed from the images corresponding to the bins indicated by red asterisks, with EF-G and P/E-position tRNA indicated with blue and red mesh, respectively.

(C) Overlapped structure shown in (B) (blue mesh) with the structure reconstructed from the images corresponding to the bins with blue asterisks (gray solid), with A/A-, P/P- and E/E-position tRNAs indicated with mesh in green, yellow, and red, respectively (one large part of the EF-G region is indicated by an arrow on the blue mesh, which helps to show that the region is occupied by no density from the gray solid structure).

(D) Overlapped structures from (C) showing different relative orientations of the 30S and 50S subunits in the two reconstructions, i.e., ratchet-like movement (gray, with EF-G; tan, without EF-G; magenta asterisk, density belonging to EF-G).

(E) Second view of the overlapped structures from (D), with arrows showing the 30S rotation and the L1 movement (gray solid, with EF-G; tan mesh, without EF-G). Although the L1 movement is rotational, it is shown as a shift to designate an inward direction of the movement.

See also [Figure S2](#) and [Movie S3](#).

containing one third of the images with the largest amplitudes (mean amplitude, 237) are indicated by blue asterisks ([Figure 2A](#)). The structures reconstructed from images in the red and blue bins clearly show two distinct conformations, with and without EF-G, respectively ([Figures 2B and 2C](#)). The structure from the midamplitude range (mean amplitude, -78) seemed to result from a mixture of different conformations with partially bound EF-G and was not further analyzed in this article.

The reconstructed structures are in excellent agreement with those obtained with other methods and the same data set

image analysis, but this movement was recovered ([Figure 2E](#)) by classifying images based on the ratchet-like motion visible in [Figures 2C and 2D](#).

These results show robustness of the proposed method to the use of a reference structure that comes from a different data set or a different composition of the studied complex (the number of tRNAs in the reference EMD-5262 structure is different from the number of tRNAs in the analyzed images, and the EF-G factor is present in some of the images but not in the EMD-5262 structure).

Experiment 3: EM Images of Pol α -B

In the absence of a structure at atomic resolution, Pol α -B was previously studied using ML3D, an EM method based on 3D maximum likelihood (Scheres et al., 2007). Negative-stain single-particle images of the complex were analyzed, which yielded a structure composed of two lobes connected by a flexible linker where the larger and smaller lobes roughly correspond to the Pol α and B subunits, respectively (Klinge et al., 2009). More precisely, ML3D revealed three different classes of images that contained 42%, 30%, and 28% of the total number of images and yielded to the reconstructions at 22.9 Å, 24.5 Å, and 25.5 Å resolutions, respectively, each corresponding to a different degree of flexion of the linker (Klinge et al., 2009). The same images (12,000 single particles) were analyzed here using HEMNMA. The structure obtained by ML3D from the largest class of images was converted into a pseudoatomic structure that was used as the reference structure to analyze images. The volume and image sizes were 64^3 voxels [voxel size, $(3.8 \text{ Å})^3$] and 64^2 pixels [pixel size, $(3.8 \text{ Å})^2$], respectively. A view of the reference conformation is shown in Figures 3A–3C (a structure with 968 pseudoatoms was obtained for the EM volume approximation error of 5%, and the SD of Gaussian functions $\sigma = 3.8$).

From 100 computed normal modes, we selected six lowest frequency non-rigid-body modes, as their collectivity degree was above 0.5 (Figure 3D). These six modes were used for the image analysis with the proposed method. The computed deformation amplitudes along the modes were then analyzed by principal-component analysis (PCA) to reduce the complexity from six to three dimensions. To visualize how the conformation changes along a mode or a principal axis (PA), a volumetric form of the pseudoatomic structure (obtained by pseudoatoms-to-volume conversion) was displaced in positive and negative directions of each of six normal modes and six PAs (Figure S3). As the PAs are linear combinations of normal modes, a contribution of different normal modes to the PAs can be observed in Figures S3G–S3L. For example, the most important PA (PA1; Figure S3G) is mainly contributed by mode 7 (Figure S3A) (a similar movement of the upper lobe is visible in Figure S3G and Figure S3A), although some other modes contribute as well (e.g., the linker is moving in Figure S3G but not in Figure S3A). Similarly, the second most important PA (PA2; Figure S3H) is mostly contributed by mode 8 (Figure S3B). The remaining PAs are more difficult to interpret visually, and their quantitative analysis showed that they were more “mixed” than PA1 and PA2 (i.e., a normal mode has a similar contribution to a PA as the other modes).

The computed deformation amplitudes are shown in Figure 3E as projections on to PA1–PA3 (each image is represented by a point), together with linear regression lines through three identified clusters of points (Figures 3E and 3I). Displacements of the volumetric form of the pseudoatomic structure along each of the three lines (Figures 3F–3H) show a movement similar to that along PA2 (Figure S3H). The main difference between the displacements along the three lines is related to the degree of an additional movement similar to that along PA1 (Figure S3G). Movies S4 and S5 show the displacement of the pseudoatomic structure and its volumetric form along the red (central) line as the highest number of images was distributed along this line.

These results clearly show a high conformational heterogeneity of the complex, which suggests that any structure ob-

tained by 3D reconstruction from these images would be of low resolution and partially inaccurate, as the actual flexibility would be limited by averaging of different conformations. Here, we present some reconstructed structures to show these effects. A structure was computed from each of the three clusters of images (Figures 3I and 3K). The overlapped structures (Figure 3K) show a movement similar to that along PA1 (Figure S3G). An overlap between the structures reconstructed along any of the three regression lines (Figure 3E) shows a movement similar to that along PA2 (Figure S3H). For example, Figure 3L shows an overlap of the structures reconstructed from three subsets (Figure 3J) of the yellow (central) cluster in Figure 3I. These reconstructed structures (Figure 3L) are similar to those shown for the displaced pseudoatomic structure (Figure 3G) along the corresponding central regression line in Figure 3E, but their resolution is low due to a variability of the incorporated conformations in the reconstruction.

As the reconstructed structures show movements similar to those along two most important PAs and these axes are mostly contributed by modes 7 and 8, we can conclude that these two modes are the most dominant modes for this data set. Interestingly enough, it has been shown that one to three modes can be sufficient to globally describe experimentally observed conformational changes (Tama et al., 2004b; Tama and Sanejouand, 2001). Our method, thus, can help to find the most dominant modes, which may be particularly useful when they are not among a few lowest frequency modes (e.g., experiment 4).

Regarding the reconstructions in Figure 3K, a similar movement was observed previously (Klinge et al., 2009), but with a smaller amplitude. To check if the smaller amplitude was due to an incorporation of different conformations into the same reconstructed structure, we designed a test involving three more reconstructions. We split the central cluster in Figure 3I into three subsets of images according to their coordinate on PA1 (each subset contained around 30% of images). Two structures were computed from two artificially mixed sets (each containing a side cluster and its neighboring central-cluster subset). The third structure was computed from the remaining central-cluster subset. Interestingly enough, these last three structures overlap quite well with the conformations obtained previously (Klinge et al., 2009) (similar movements with similar amplitudes can be observed in Figure 3M). The smaller amplitude of the movement observed previously can thus be explained by a heterogeneity of conformations incorporated in the reconstructed average structures (the flexibility could not be fully explored with ML3D using a small number of classes, and the full exploration would require a much larger number of classes and images; Klinge et al., 2009). HEMNMA described more extensively the conformational heterogeneity that was detected by ML3D using the same data set. For example, it offered the possibility to additionally visualize the changes of the linker length coupled with the changes of the relative distance between the lobes (movements reflecting mainly displacements along PA2 (Figures 3G and 3L; Figure S3H)).

Experiment 4: EM Images of TBSV

Swelling of many icosahedral plant viruses has been observed upon changes of pH and EDTA chelation of divalent cations on

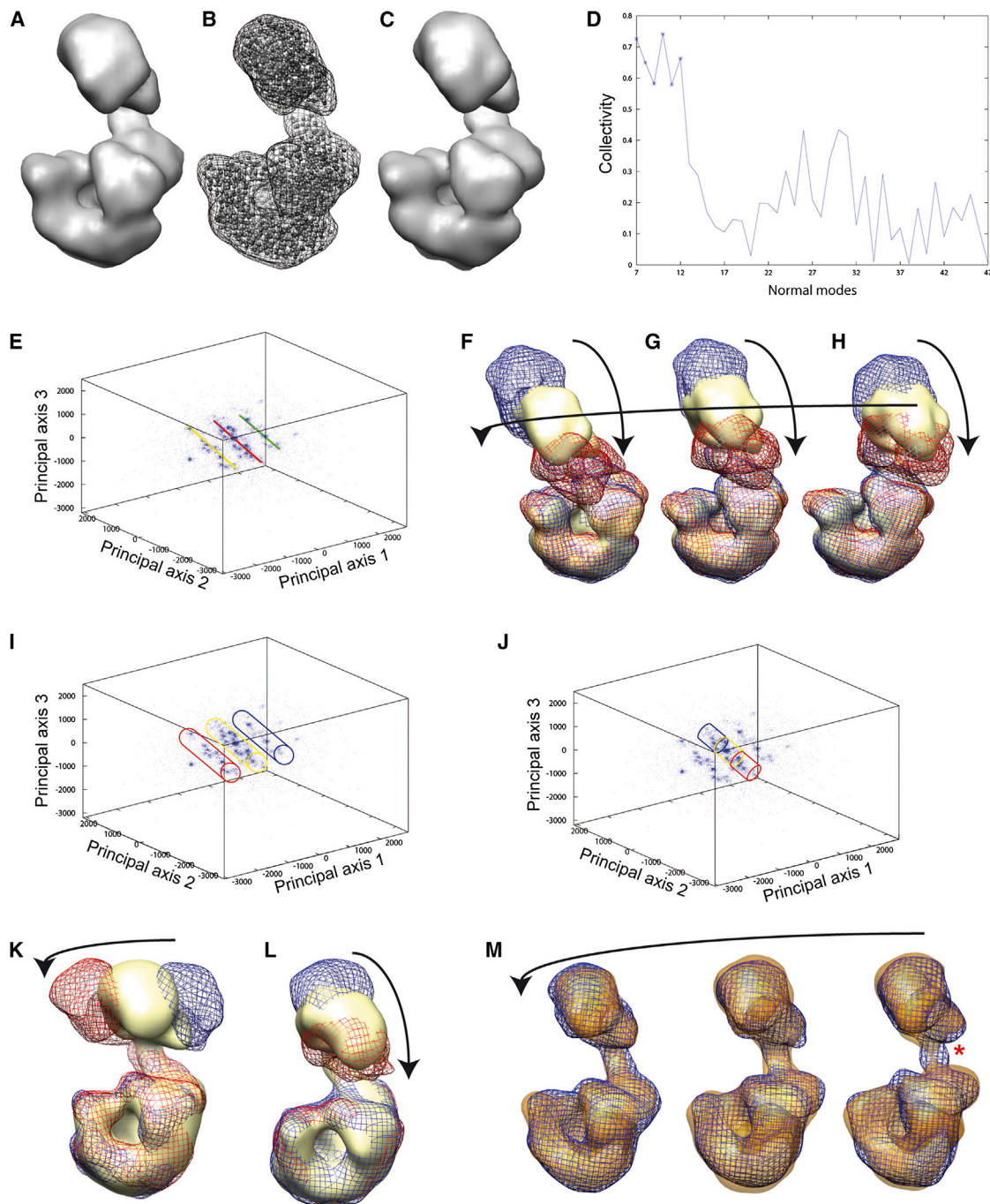


Figure 3. Experiment with EM Images of Pol α -B

(A) EM volume obtained by ML3D (from the largest class of images).

(B) Pseudoatomic structure (spheres) from the volume in (A) overlapped with the volume (mesh).

(C) Approximation of the volume shown in (A) by the pseudoatomic structure shown in (B).

(D) Collectivity degree of normal modes computed for the pseudoatomic structure (the first six modes related to rigid-body movements are not shown; for better visibility, only 40 modes are shown; the asterisks denote the modes with the collectivity degree above 0.5).

(E) Computed deformation amplitudes projected on to the three most important PAs (the points denote single-particle images and PA1, the most important PA), and linear regression lines (yellow, red, and green) through three clusters; see also (I).

(F–H) Volumetric form of the pseudoatomic structure displaced along the lines shown in (E) [(F), (G), and (H) for the yellow, red, and green lines, respectively], with the overlapped structures from two endpoints of the line (blue mesh and red mesh) and the midpoint (yellow solid).

(I) Clusters of images denoted by cylinders.

(J) Three equal-sized subsets of the yellow (central) cluster in (I).

(legend continued on next page)

the interface between subunits (Ca^{2+} or Mg^{2+} that appear to play a critical role in virion stability), but the swelling mechanism is still not well understood (Witz and Brown, 2001). In a previous EM study of the TBSV swelling, a structure of the native (compact) TBSV form was obtained at 13 Å resolution, and a continuous particle size variation was found on micrographs of the virus after its dialysis in Tris-EDTA buffer (Aramayo et al., 2005). Also, a structure of the most swollen form was obtained at 19 Å resolution by an iterative procedure whose each iteration comprised image alignment, sorting of the largest particles (based on a threshold cross-correlation with the last reference structure), and 3D reconstruction to compute a new reference structure for the next iteration (Aramayo et al., 2005). HEMNMA was used here to analyze a set of 7,000 TBSV images.

The EM structure and image sizes were reduced to 128^3 voxels and 128^2 pixels, respectively [pixel size, $(3.2 \text{ Å})^2$]. The pseudoatomic structure for NMA was obtained from the structure of the compact form (6,096 pseudoatoms for the volume approximation error of 5%, and for the SD of Gaussian functions $\sigma = 3.2$) (Figure 4A). From 100 computed normal modes, image analysis was conducted using four lowest frequency non-rigid-body modes whose collectivity degree was above 0.75 (modes 13, 24, 40, and 77, marked by asterisks in Figure 4B). No information was used about the potential symmetry of the conformations during the image analysis, and the used modes allowed for asymmetric geometrical transformations. In fact, only one of the used modes deforms the structure while perfectly preserving its symmetry (mode 24, which is related to the radially symmetric expansion). The other three modes result in asymmetric changes. The deformation amplitudes computed through image analysis for each of the four modes were then mapped on to the 3D PCA space (Figure 4C). The data were fitted with a linear regression line in the PCA space, and the reference structure was displaced along this line (blue asterisks in Figure 4C; Movie S6). Also, the data were split into four classes according to the data coordinate on PA1 (four equal segments of the axis), and 3D reconstruction was conducted from the images in each class (Figure 4D). The colors red, magenta, green, and yellow in Figures 4C and 4D correspond to the four classes with 21%, 25%, 24%, and 30% of all images, respectively. Although the image analysis was conducted with no a priori information about the conformational symmetry, the reconstructed structures were icosahedrally symmetrized to reduce noise, as it was done in the previous study (Aramayo et al., 2005). The difference between the diameter of the structure from the most swollen particles (yellow in Figure 4D) and that of the compact-form structure (gray in Figure 4D) is about 4 nm, which is consistent with the previously reported differences (Aramayo et al., 2005; Krüse et al., 1982). The difference in the diameter of the structure from the most and least swollen particles (yellow and red in Figure 4D, respectively) is about 3.2 nm. This indicates that the diameter of the least swollen particles in the analyzed images

is larger than that of the compact-form particle and that the particles were all more or less swollen in the analyzed samples.

The conformational change can be mainly described as a combination of an increase in the radius of the particle (increase in the negative direction of PA1), a rotation of the subunits around the 5-fold symmetry axis, and an increase in the size of the intersubunit space at the subunits quasi-trimer (central part of the capsid in Figure 4D and Movie S6). These results are consistent with the models of the TBSV conformational change proposed in the literature (Aramayo et al., 2005; Krüse et al., 1982; Robinson and Harrison, 1982). Most important, the visualization of the full conformational distribution revealed many intermediate states of the change, which describes more extensively the conformational heterogeneity that was detected previously (Aramayo et al., 2005).

DISCUSSION

This article describes the method HEMNMA that allows analysis of the dynamics of large macromolecular complexes by experimental EM image analysis of possible conformational-change directions predicted by NMA. HEMNMA is particularly useful in the cases in which the reference structure at atomic resolution is unavailable but a structure can be obtained by EM.

Full Dynamics Can Be Studied

HEMNMA is ideally suited to study full dynamics, as it does not use classification during the image analysis. The classification of the computed deformation amplitudes can be performed at the last step to compute structures with similar conformations. We showed that the conformational transition conveyed by the reconstructed structures is consistent with the one conveyed by the reference structure displaced along the trajectories fitted through the computed deformation amplitudes. However, the full conformational distribution conveys additional information that is not contained in the reconstructions obtained from the classes. Therefore, when flexibility is described by a large range of conformations, a clear advantage of our method is in looking for a full range of conformations, contrary to the standard methods that only look for discrete and limited number of conformational states.

HEMNMA Robustness

In the ribosome experiments (experiments 1 and 2), we analyzed images with the proposed method using only the mode related to ratchet-like motion to see whether we could recover both ratchet-like and L1-stalk motions. The motions were successfully recovered, and robustness of the method was shown to different realistic factors (e.g., different SNRs, different CTFs, different resolutions and compositions of the reference structure with respect to those of the imaged complexes).

(K) Structures reconstructed from the clusters in (I) (the color of the structure is that of the cluster).

(L) Structures reconstructed from the subsets in (J), with the same coloring as that used in (K).

(M) Structures from mixed subsets (blue mesh, see the text for more details) overlapped with the structures obtained with ML3D (orange solid). The red asterisk denotes an absence of the complete mass in the linker region obtained by ML3D, which is an evidence of the conformational heterogeneity of the image set used to compute this structure. Arrows show movements.

See also Figure S3 and Movies S4 and S5.

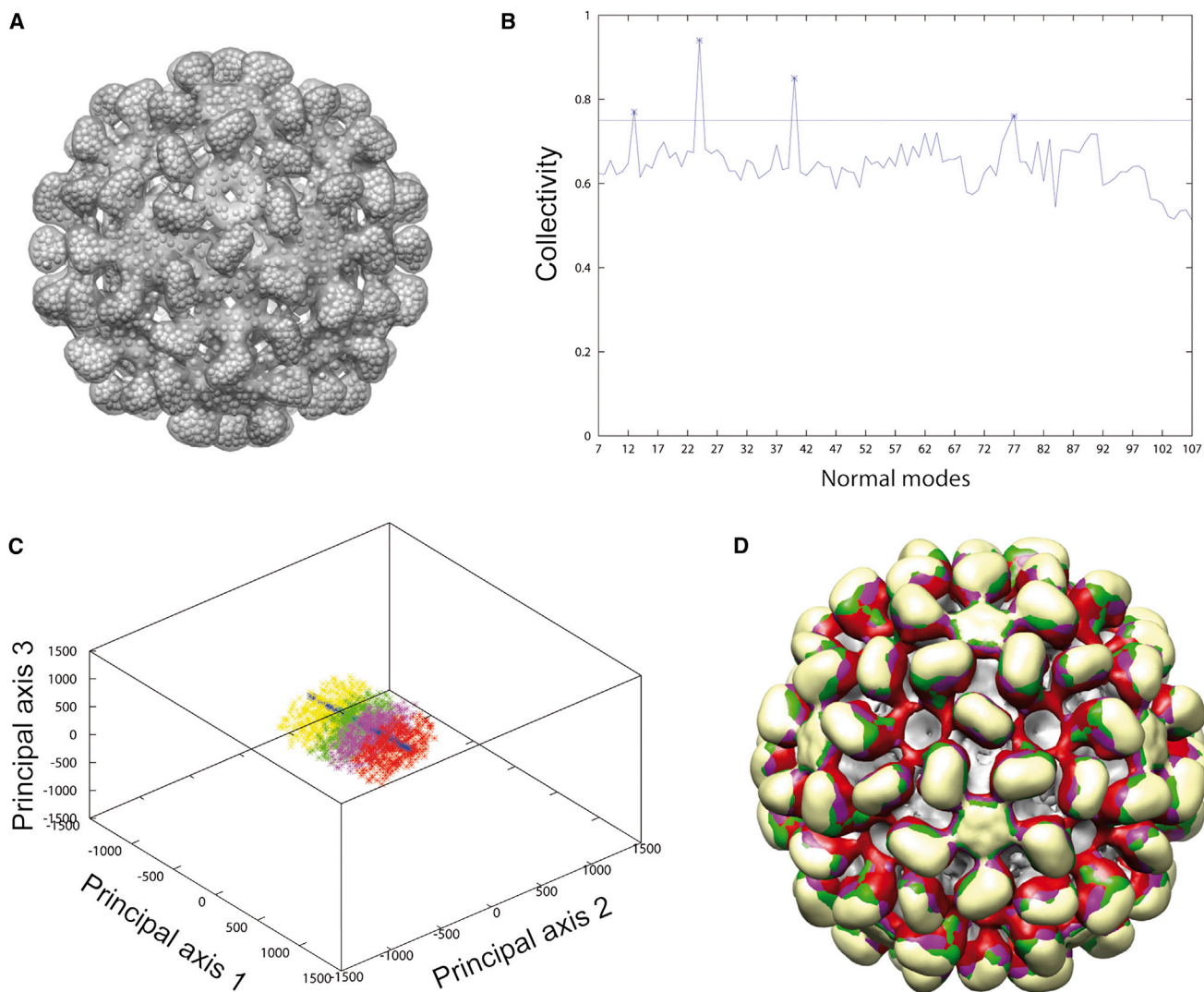


Figure 4. Experiment with EM Images of TBSV

(A) Pseudoatomic structure (spheres) computed using a compact-form TBSV structure from a previous EM study (the EM structure [Aramayo et al., 2005] is shown in transparent gray).

(B) Collectivity degree of the pseudoatomic normal modes. The first six modes related to rigid-body movements are not shown; the horizontal line corresponds to the collectivity of 0.75; the asterisks denote the modes with the collectivity degree above 0.75.

(C) Computed deformation amplitudes projected on to the three most important PAs (the points denote single-particle images and PA1, the most important PA), the linear regression line (blue), and four groups of images (red, magenta, green, and yellow) determined by splitting the full range of the data coordinates on the first PA into four equal-length intervals.

(D) Overlapped structures reconstructed from the four groups of images shown in (C). The volumes are indicated with the color of the corresponding group in (C). See also [Movie S6](#).

Complexes of Different Composition Can Be Studied

The experiment with EM images of 70S ribosome (experiment 2) was principally used to show that the method can work with different compositions of complexes in the same sample (with EF-G, without EF-G, and with partially bound EF-G) that, at the same time, produce different conformations of the complexes. We refer to this kind of conformational variability to as “discrete-type conformational changes,” because the sample is prepared so that a small number of conformational classes are expected and the problem can be solved by classical (“discrete”) methods. This experiment shows that, in the case

of such “discrete-type” problems, our method can reproduce the results obtained using classical methods and the same data (Elad et al., 2008; Scheres et al., 2007). The functional significance of these results was reported in the first paper in which this data set was analyzed (Scheres et al., 2007).

Functional Significance of Observed Motions

The experiments with Pol α -B and TBSV highlight the main characteristics of the methodology, showing that it can be used to explore truly continuous-type conformational changes where the complex cannot be stabilized in one or a few particular

conformations. They also showed how the method works when no *a priori* knowledge about the conformational change is used for the selection of normal modes. The method can find contributions of the used modes to the PAs as well as their contributions to the principal trajectories that determine the most dominant motions. Moreover, the methodology is particularly useful in identifying the most dominant modes when they are not among a few lowest frequency modes (e.g., as in the TBSV case). These experiments are also interesting as they show that the method allows studying conformational pathways that may actually exist but may be difficult to solve by X-ray crystallography or classical EM methods that rely on a predefined initial number of classes.

Pol α -B is arranged as an elongated structure organized in two lobes connected by a flexible linker. This quaternary organization has been observed for other eukaryotic replicative polymerases such as DNA polymerase ϵ (Asturias et al., 2006) and DNA polymerase δ (Jain et al., 2009), which are responsible for the replication of the leading and lagging strands, respectively. In all these replicative complexes, the flexibility between the different lobes has been identified as a functional characteristic that allows the accommodation of different conformations of DNA during its replication and facilitates the interactions of these complexes with other components of the replication machinery. An EM analysis of the structure of the complete eukaryotic primosome, consisting of Pol α , the B subunit, and the two components of the primase (PriL and PriS) revealed a very flexible connection between the two lobes and that this complex was capable of adopting a large range of different conformations (Núñez-Ramírez et al., 2011). This flexibility was postulated as a mechanism for the transfer of the DNA substrate from the primase catalytic site placed in one of the lobes to the DNA polymerase catalytic site located in the other lobe. It is surprising that ML3D image classification of Pol α -B, a subcomplex, comprising only two of the components of the primosome, revealed a limited flexibility between the catalytic core of Pol α (the big lobe) and the B-CTD platform (the small lobe) (Klinge et al., 2009). The distinct behavior of the full primosome complex and the Pol α -B subcomplex was difficult to rationalize. Now, the analysis of the same data set with HEMNMA has revealed that the high degree of flexibility showed in the full complex is also present in the Pol α -B subcomplex. Intermediate conformations of the Pol α -B subcomplex were undetected by ML3D, probably because of the difficulty in assigning a large range of conformations to a discrete number of solutions. In addition, the classical refinement-based methods use an initial template and typically discard images that poorly correlate with the average 3D model obtained during the template refinement. Thus, if a dominant conformation was used to build the initial reference, we suspect that the particles corresponding to distant conformations would be removed during the refinement by classical methods. HEMNMA has provided a more precise description of the continuous range of conformations of Pol α -B complexes, allowing a better understanding of their structure-function relationship.

The visualization of the full conformational distribution of the TBSV helped to reveal many intermediate conformations that were previously hypothesized but impossible to visualize (Aramayo et al., 2005). Indeed, many intermediate conformations might explain a difficult crystallization of such swelling viruses (e.g., crystallography produced a swollen TBSV structure at res-

olution of only 8 Å; Robinson and Harrison, 1982). The Ca^{2+} ions are known to be present between the subunits forming the TBSV capsid in order to strengthen the links between them. Many different intermediate conformations may be explained by an incomplete removal of these ions by EDTA (sample preparation is described by Aramayo et al., 2005) that could lead to a partial opening of the viruses, partially held by remaining Ca^{2+} ions. Similar conformations may exist as mechanical intermediate states permitting the release of the viral RNA within the host cell. The genome of the TBSV consists of a single RNA filament shifting toward the capsid in an environment depleted of Ca^{2+} ions, and such intermediate conformations could be enough for the release of the single RNA filament *in vivo*.

Thus, HEMNMA seems highly suitable for macromolecular complexes displaying a continuous range of conformations. This may be the case of a large number of complexes in biology. In these conditions, HEMNMA seems significantly more powerful than the methods with a predefined initial number of classes, which would require larger data sets and larger computing times to evaluate the presence of a large number of possible conformers. However, the classification-based methods can be combined with HEMNMA to better analyze overall heterogeneity (e.g., by providing one or several reference structures as in the Pol α -B case).

EXPERIMENTAL PROCEDURES

Pseudoatomic Structures from EM Volumes

To obtain a reduced EM density representation, our recently published method for volume-to-pseudoatoms conversion (Nogales-Cadenas et al., 2013) was used as it allows controlling the volume approximation error and, thus, controlling quality of the projections for the elastic projection-matching-based image alignment. The used conversion method represents a volume by a collection of Gaussian functions whose number, positions, and amplitudes are adjusted for a given volume approximation error and a given Gaussian standard deviation. The Gaussian functions are referred to as pseudo-atoms although their positions do not generally coincide with the atomic positions, and the structure with pseudo-atoms is referred to as pseudo-atomic structure (Nogales-Cadenas et al., 2013).

NMA

To compute normal modes, a simplified elastic network representation of the potential energy function (Tirion, 1996) was used, as implemented on the Elnemo web server (Suhre and Sanejouand, 2004) for atomic structures and the 3DEM Loupe web server (Nogales-Cadenas et al., 2013) for EM structures. Before computing normal modes, 3DEM Loupe converts the input EM volume into a pseudoatomic structure that can be directly used as the reference structure for the method proposed here. The coordinates of the normal mode vectors are computed in angstroms, while the deformation amplitudes have no units. Six lowest frequency modes (modes 1–6) are not used as they are related to rigid-body movements.

SUPPLEMENTAL INFORMATION

Supplemental Information includes Supplemental Experimental Procedures, three figures, one table, and six movies and can be found with this article online at <http://dx.doi.org/10.1016/j.str.2014.01.004>.

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