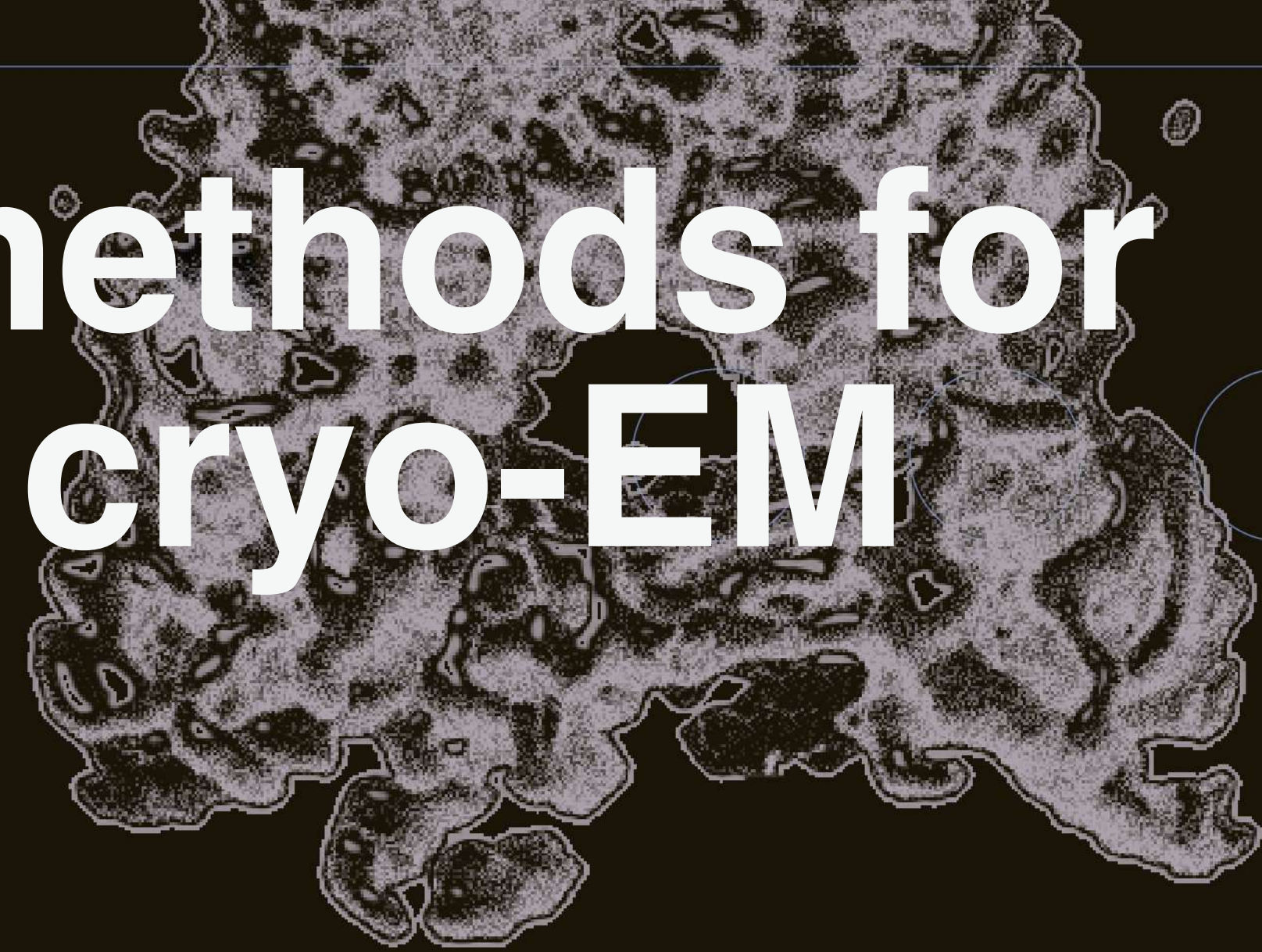
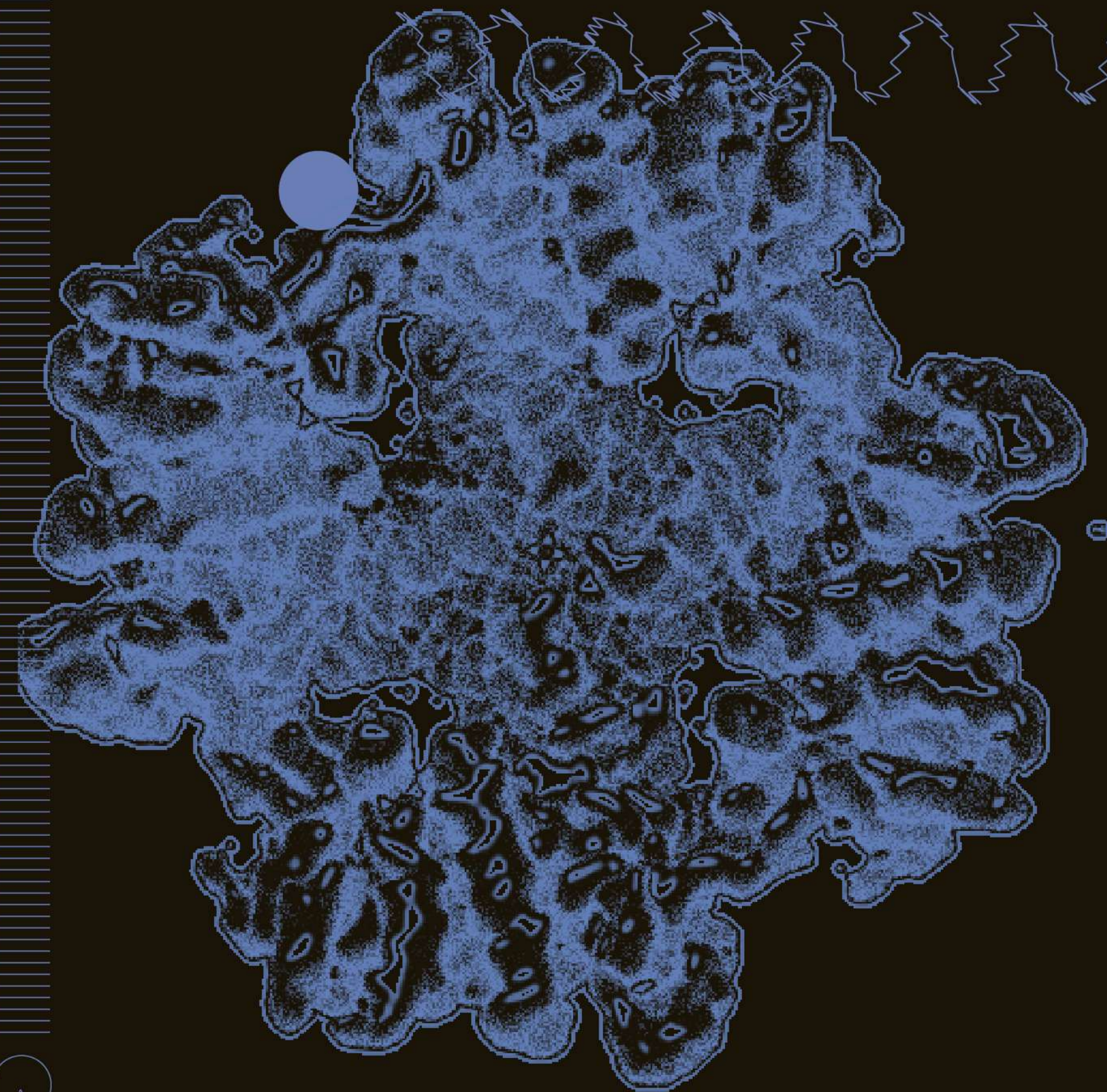
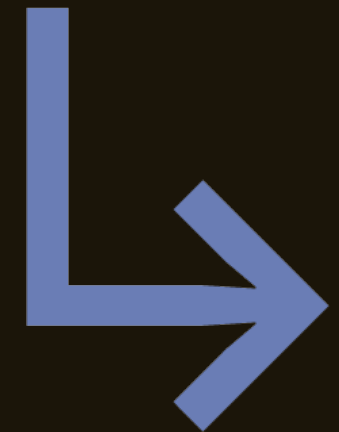


ML methods for heterogeneity in SPA →

Sonya M. Hanson
Structural and Molecular Biophysics, CCB/CCM
Flatiron Institute, Simons Foundation, NYC

NYSBC-NCCAT SPA short course 2024

Computational methods for heterogeneity in cryo-EM



ML methods for heterogeneity in cryo-EM

- PCA
- ManifoldEM
- cryoDRGN
- GMM in eman2
- cryoSPARC/RELION
- ... and more!

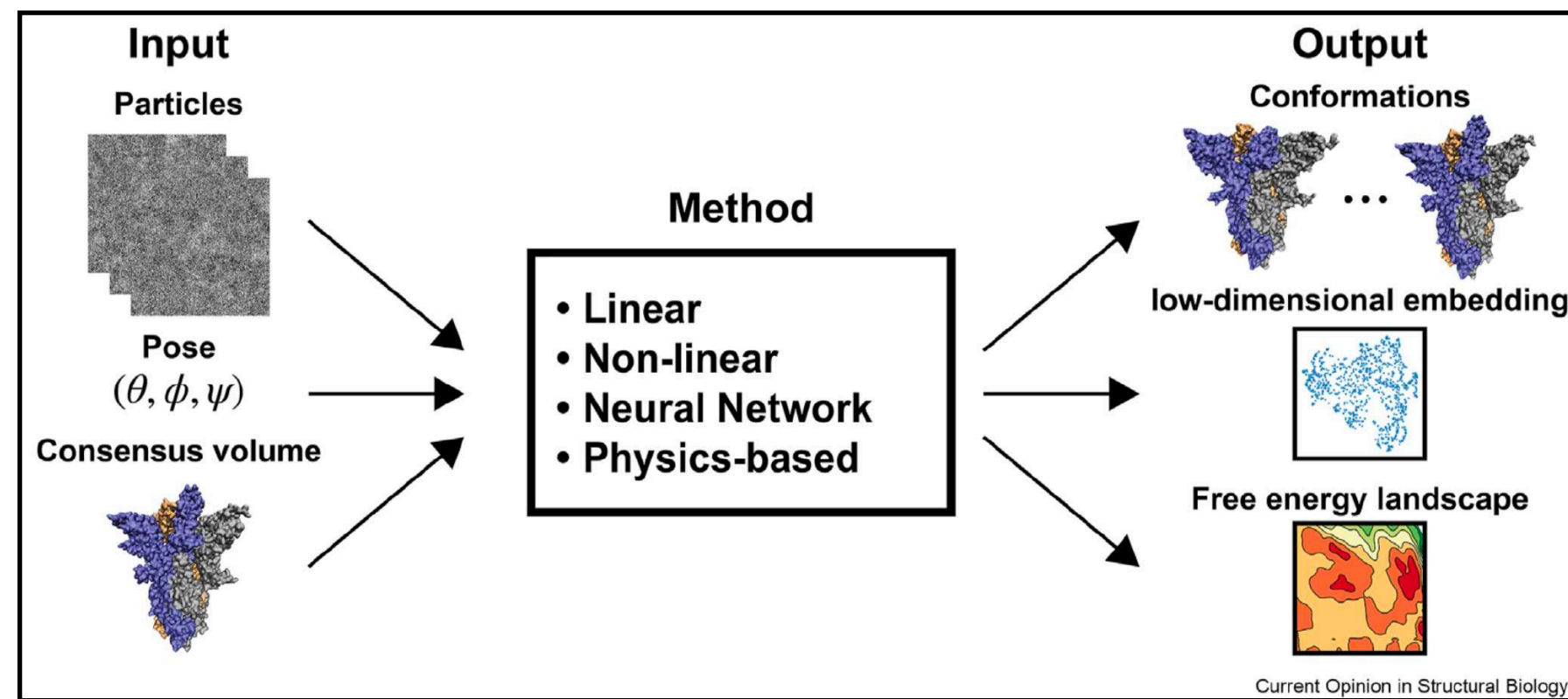
Practical considerations when using continuous heterogeneity methods.

Combining cryo-EM and Molecular Dynamics
Assessing the performance of continuous heterogeneity methods.

A couple useful reviews to check out.

Conformational heterogeneity and probability distributions from single-particle cryo-electron microscopy

Wai Shing Tang¹, Ellen D. Zhong³, Sonya M. Hanson^{1,2}, Erik H. Thiede¹ and Pilar Cossio^{1,2}



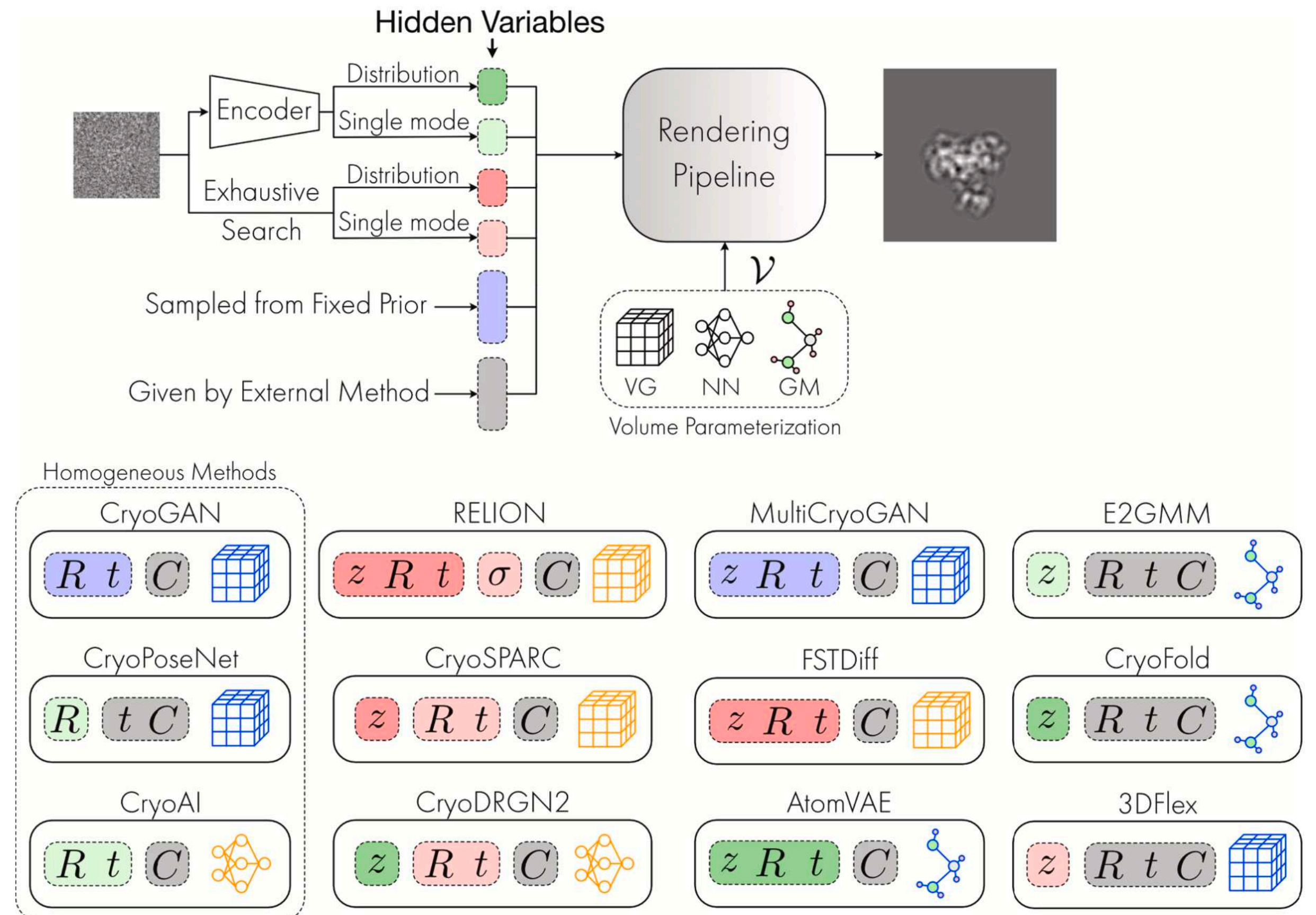
www.sciencedirect.com

Current Opinion in Structural Biology 2023, 81:102626

Journal of Structural Biology 214 (2022) 107920

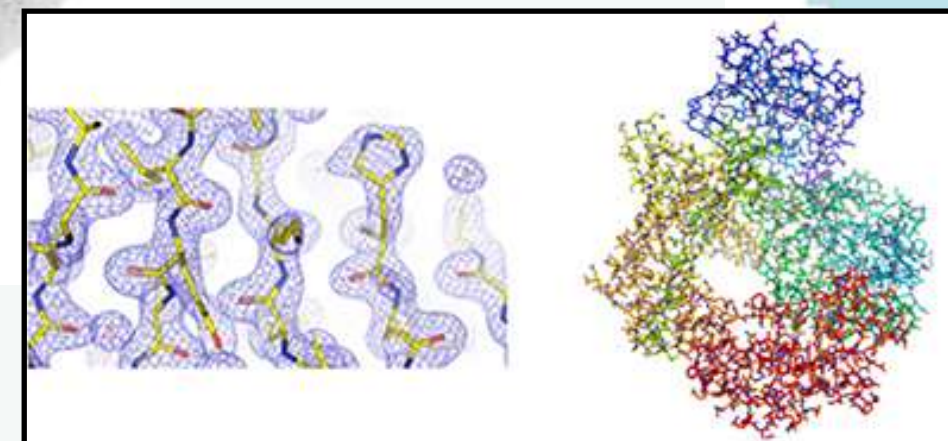
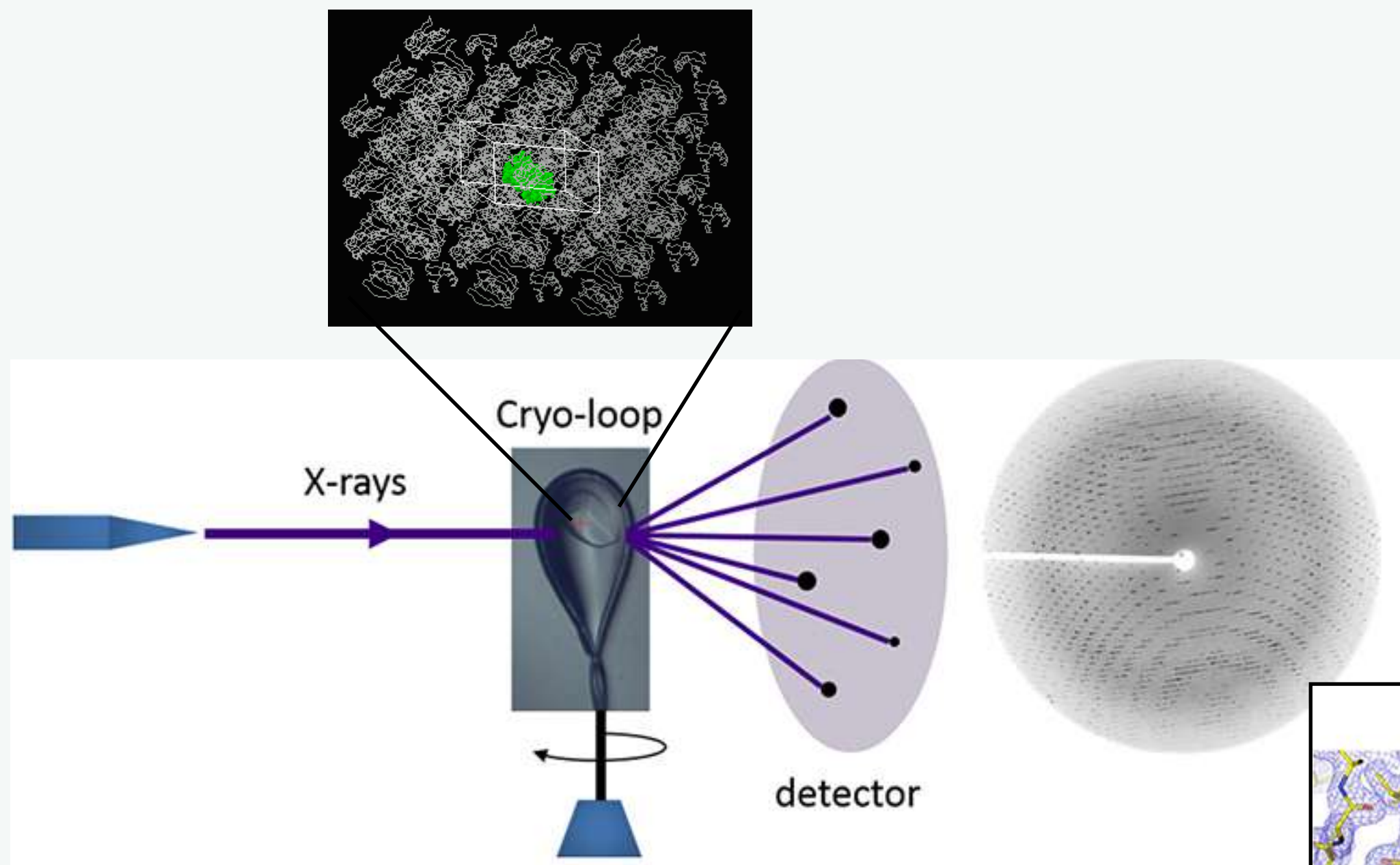
Deep generative modeling for volume reconstruction in cryo-electron microscopy

Claire Donnat^{a,1}, Axel Levy^{b,c}, Frédéric Poitevin^c, Ellen D. Zhong^d, Nina Miolane^{e,1}

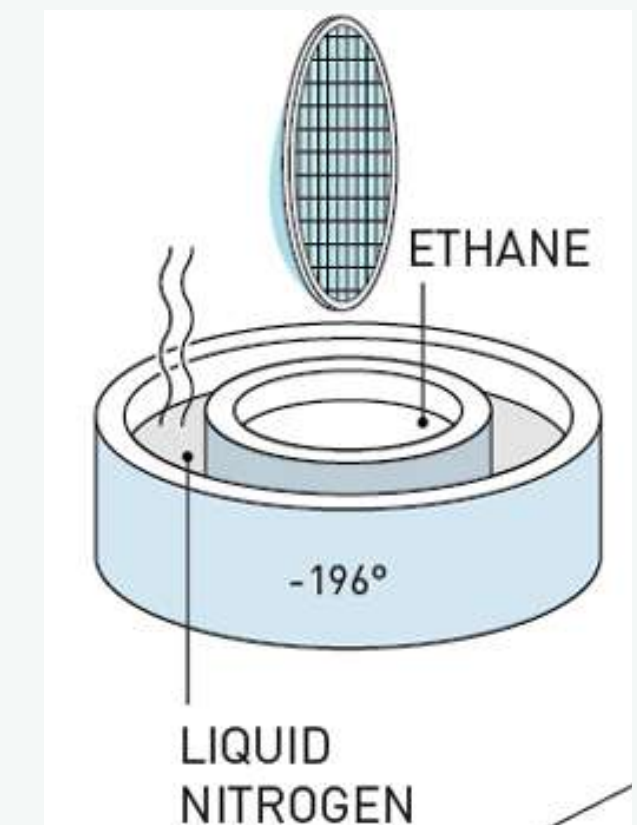
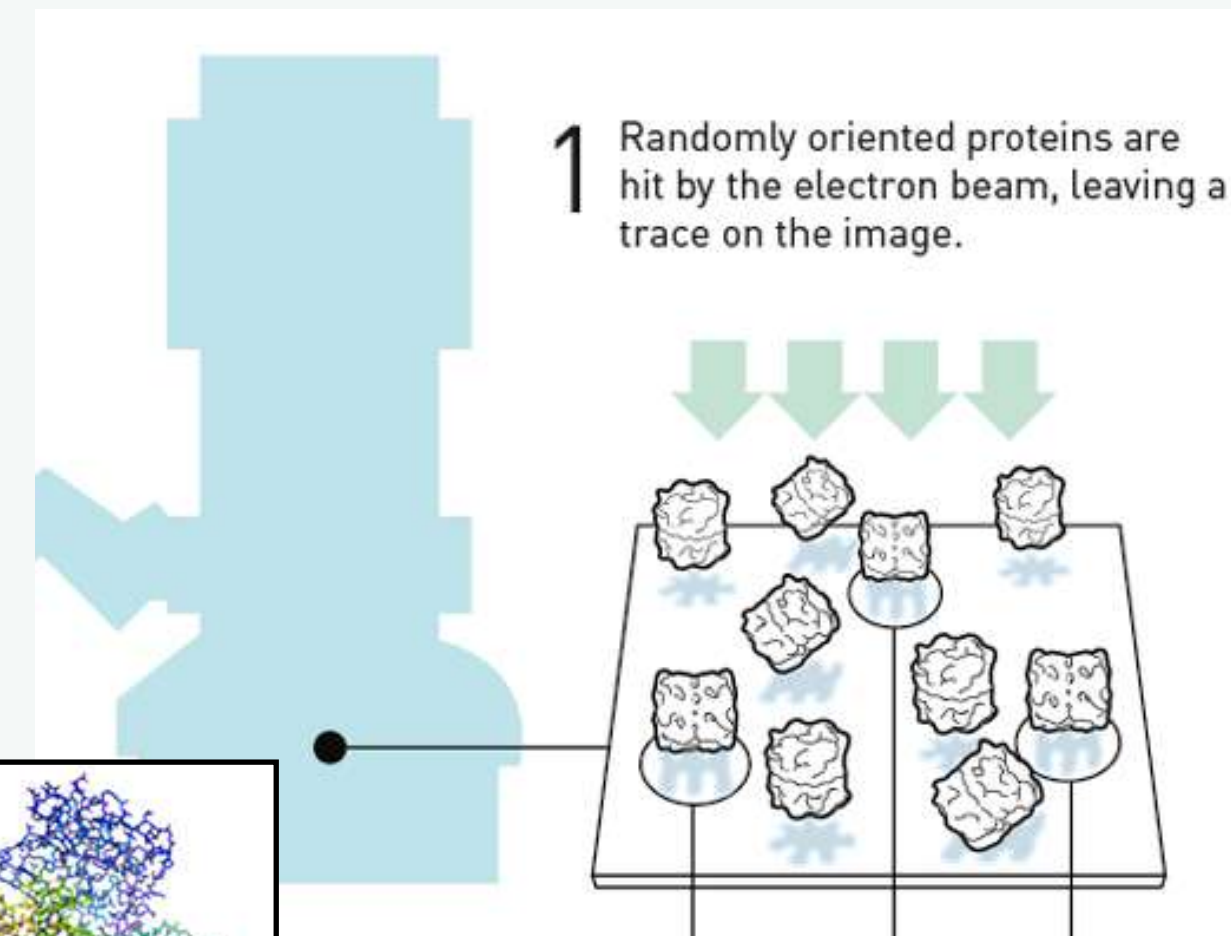


A key advantage of cryo-EM not yet fully exploited!

X-ray crystallography



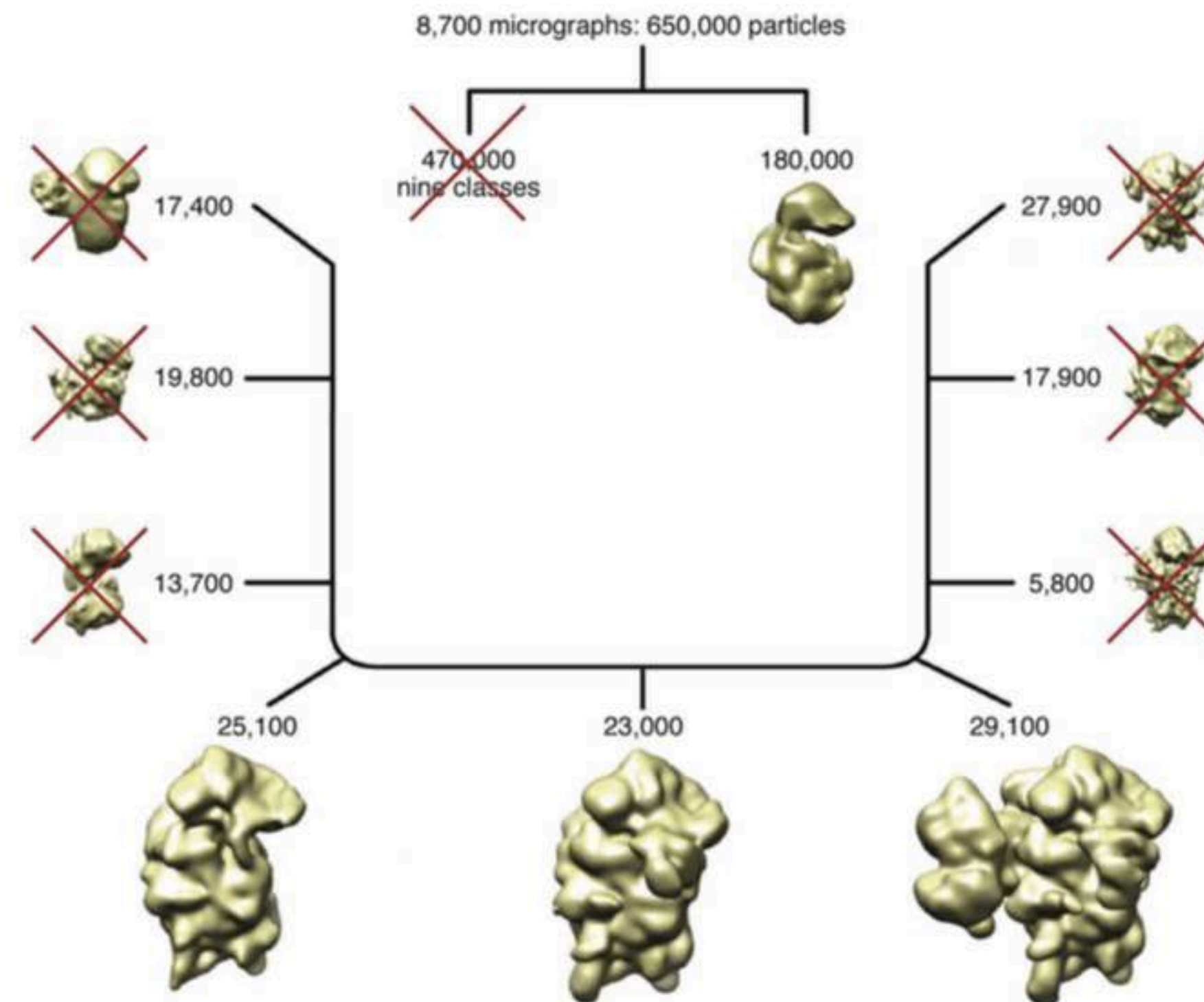
cryo-EM



A key advantage of cryo-EM still to be fully exploited!

Traditional 3D classification methods are imperfect.

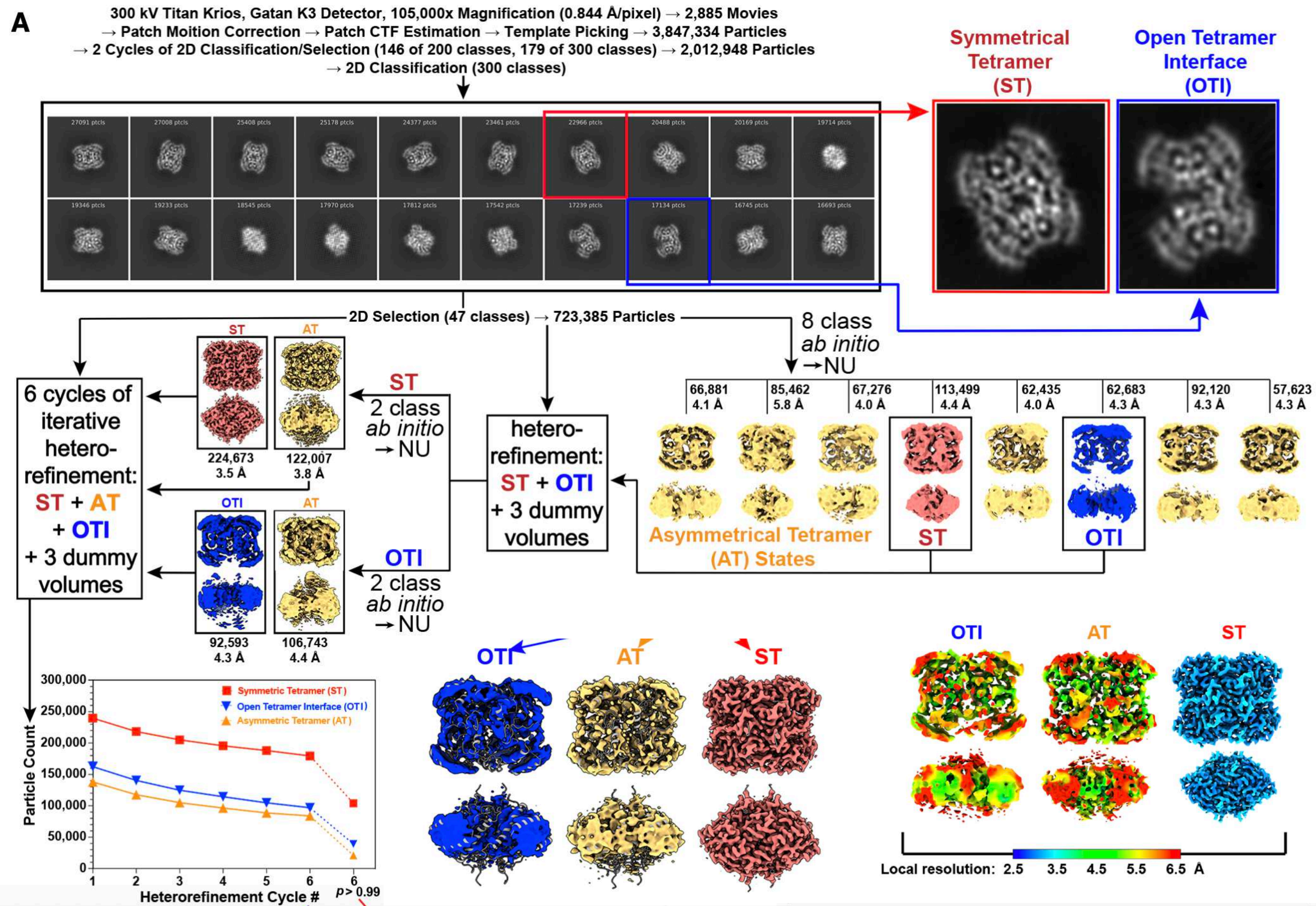
- Require a user-defined number of discrete states.
- Average out any other heterogeneity.



Two rounds of 3D classification:
10 classes, then 9 classes.

Frank & Ourmazd,
Methods (2016)

Discrete classes in cryo-EM lose information.

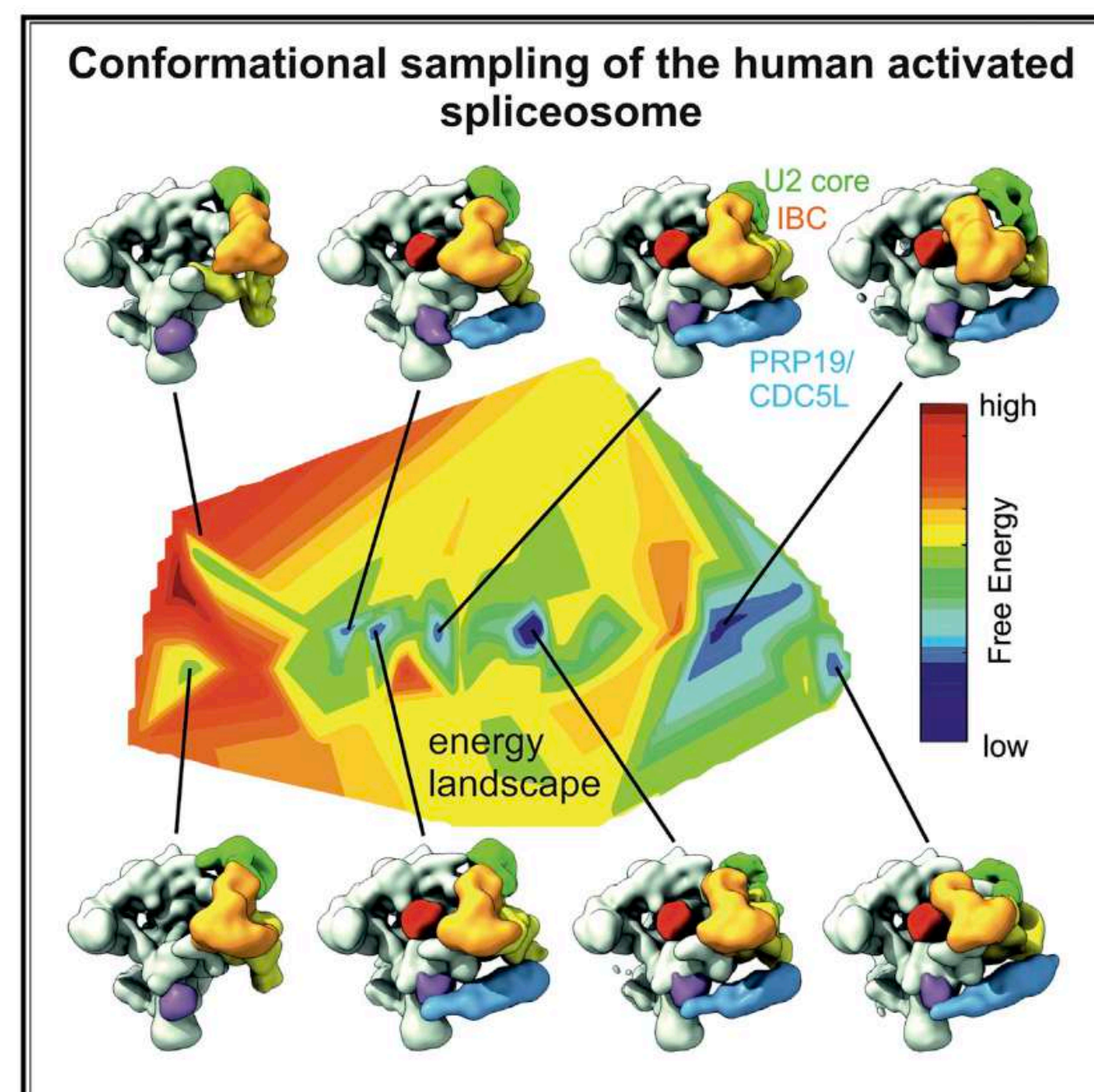


Principal component analysis (PCA) for continuous heterogeneity in SPA.

Cell

Structure and Conformational Dynamics of the Human Spliceosomal B^{act} Complex

Graphical Abstract



Authors

David Haselbach, Ilya Komarov, Dmitry E. Agafonov, ..., Berthold Kastner, Reinhard Lührmann, Holger Stark

Correspondence

reinhard.luehmann@mpi-bpc.mpg.de (R.L.), hstark1@gwdg.de (H.S.)

In Brief

A new approach to analyzing cryo-EM data reports on conformational dynamics in the human spliceosome.

Haselbach et al., 2018, Cell 172, 454–464
January 25, 2018 © 2018 Elsevier Inc.
<https://doi.org/10.1016/j.cell.2018.01.010>

Article



STRUCTURAL
BIOLOGY

ISSN 2059-7983

Acta Cryst. (2021). D77, 835–839

Principal component analysis is limited to low-resolution analysis in cryoEM

Carlos Oscar S. Sorzano* and Jose Maria Carazo

New Results

Posted November 01, 2023.

Follow this preprint

A Bayesian Framework for Cryo-EM Heterogeneity Analysis using Regularized Covariance Estimation

Marc Aurèle Gilles, Amit Singer

doi: <https://doi.org/10.1101/2023.10.28.564422>

bioRxiv
THE PREPRINT SERVER FOR BIOLOGY

ManifoldEM

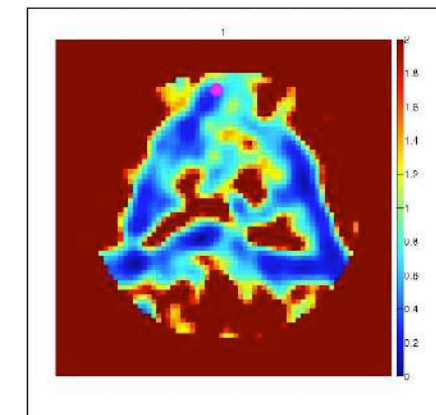
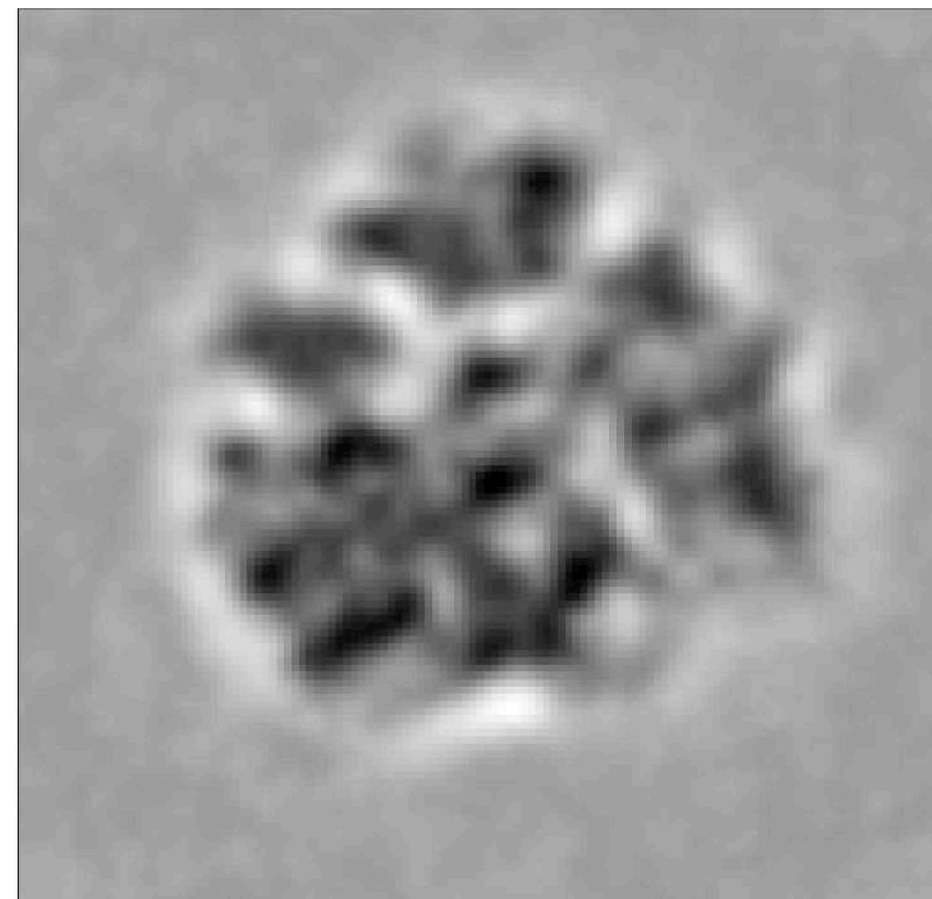
PNAS

Trajectories of the ribosome as a Brownian nanomachine

Ali Dashti^{a,1}, Peter Schwander^{a,1}, Robert Langlois^b, Russell Fung^a, Wen Li^b, Ahmad Hosseinizadeh^a, Hstau Y. Liao^b, Jesper Pallesen^{c,2}, Gyanesh Sharma^{b,3}, Vera A. Stupina^d, Anne E. Simon^d, Jonathan D. Dinman^d, Joachim Frank^{b,c,4}, and Abbas Ourmazd^{a,1,4}

^aDepartment of Physics, University of Wisconsin, Milwaukee, WI 53211; ^bDepartment of Biochemistry and Molecular Biophysics, and ^cHoward Hughes Medical Institute, Columbia University, New York, NY 10032; and ^dDepartment of Cell Biology and Molecular Genetics, University of Maryland, College Park, MD 20742

Contributed by Joachim Frank, October 8, 2014 (sent for review September 10, 2014)



Dashti et al, *PNAS* (2014).

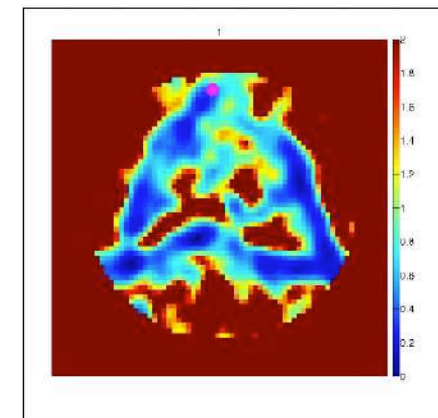
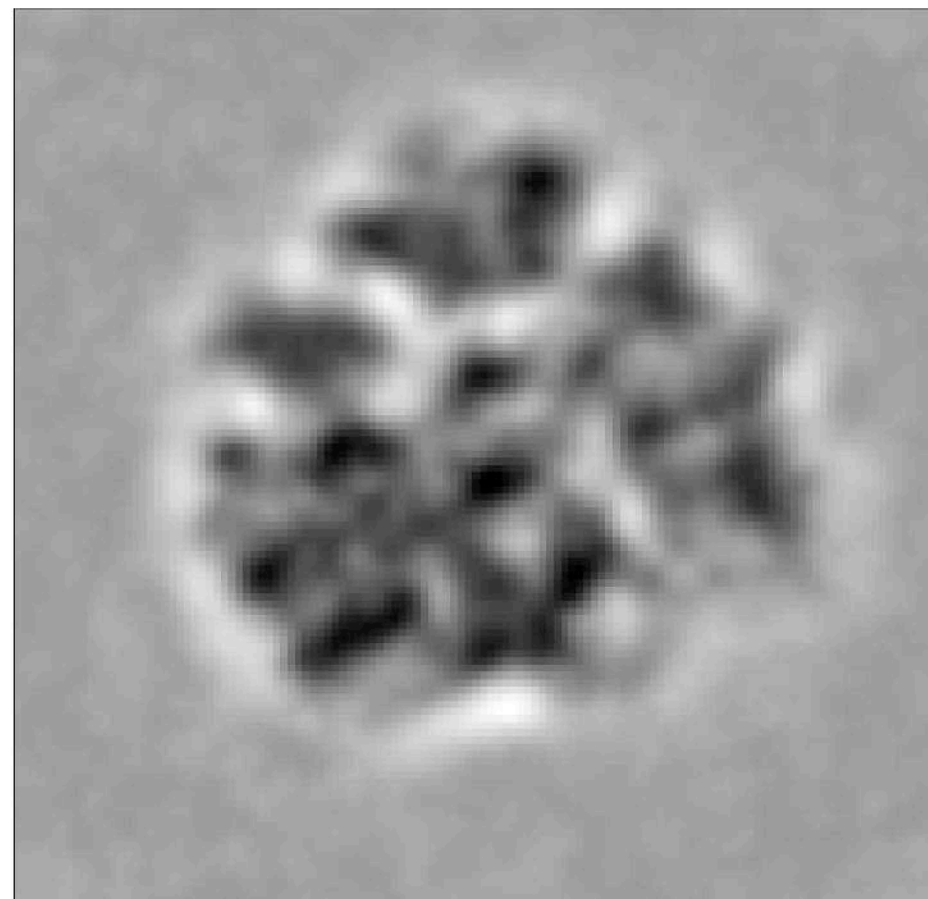
ManifoldEM

Trajectories of the ribosome as a Brownian nanomachine

Ali Dashti^{a,1}, Peter Schwander^{a,1}, Robert Langlois^b, Russell Fung^a, Wen Li^b, Ahmad Hosseinizadeh^a, Hstau Y. Liao^b, Jesper Pallesen^{c,2}, Gyanesh Sharma^{b,3}, Vera A. Stupina^d, Anne E. Simon^d, Jonathan D. Dinman^d, Joachim Frank^{b,c,4}, and Abbas Ourmazd^{a,1,4}

^aDepartment of Physics, University of Wisconsin, Milwaukee, WI 53211; ^bDepartment of Biochemistry and Molecular Biophysics, and ^cHoward Hughes Medical Institute, Columbia University, New York, NY 10032; and ^dDepartment of Cell Biology and Molecular Genetics, University of Maryland, College Park, MD 20742

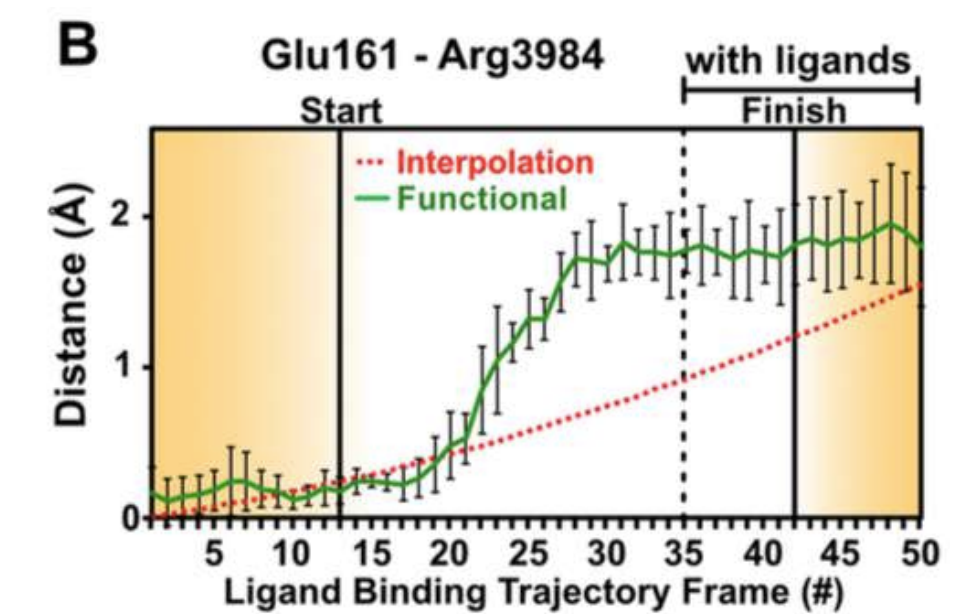
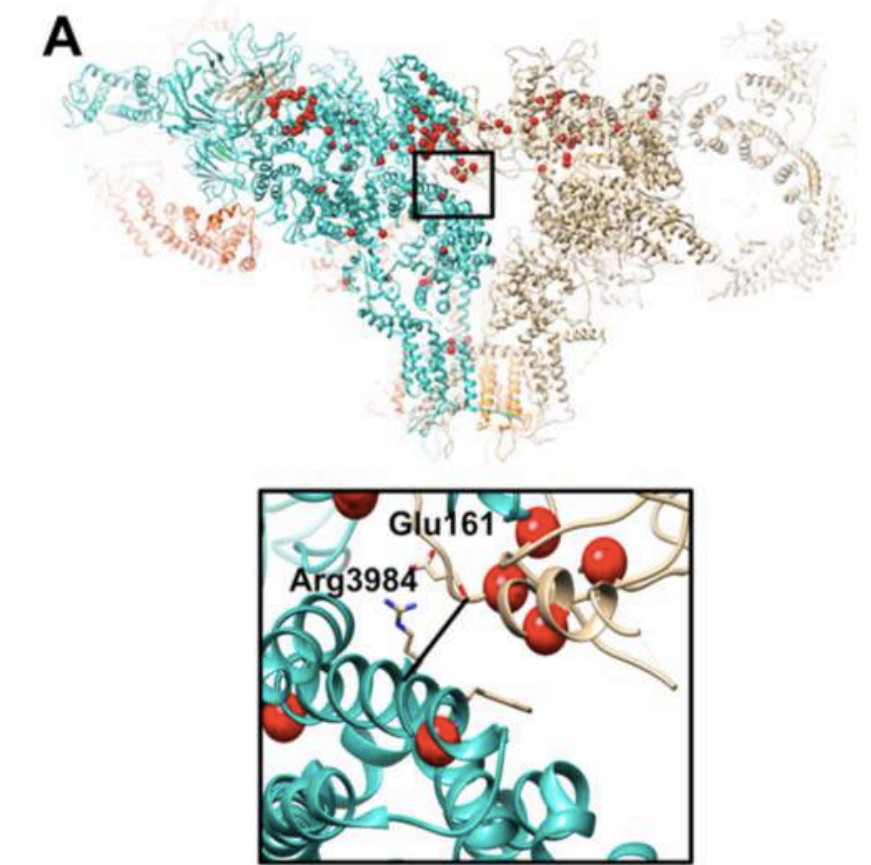
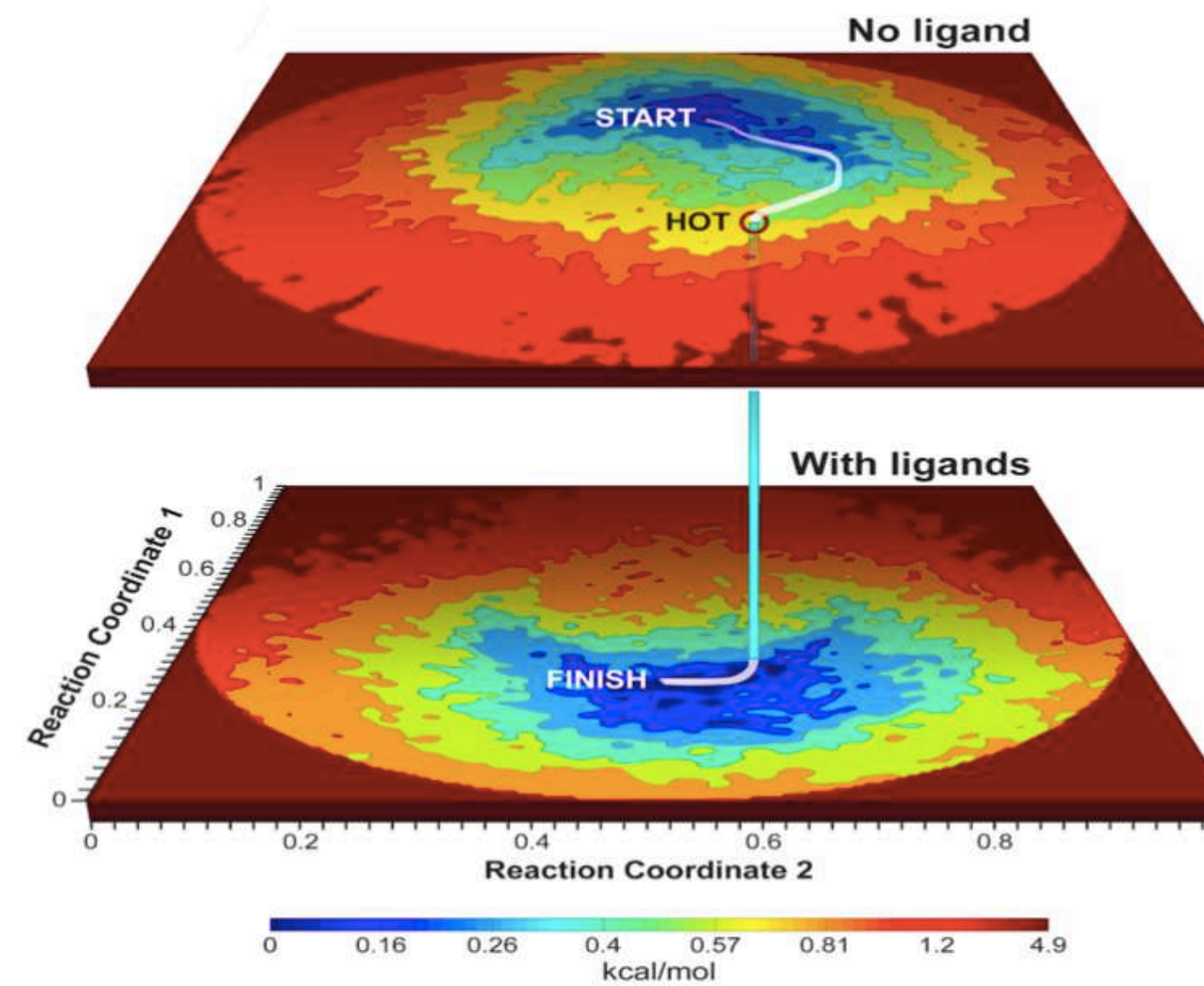
Contributed by Joachim Frank, October 8, 2014 (sent for review September 10, 2014)



Dashti et al, *PNAS* (2014).

Retrieving functional pathways of biomolecules from single-particle snapshots

Ali Dashti^{1,9}, Ghoncheh Mashayekhi^{1,9}, Mrinal Shekhar^{2,3}, Danya Ben Hail⁴, Salah Salah^{4,5,6}, Peter Schwander¹, Amedee des Georges^{4,5,6}, Abhishek Singharoy³, Joachim Frank^{7,8} & Abbas Ourmazd¹

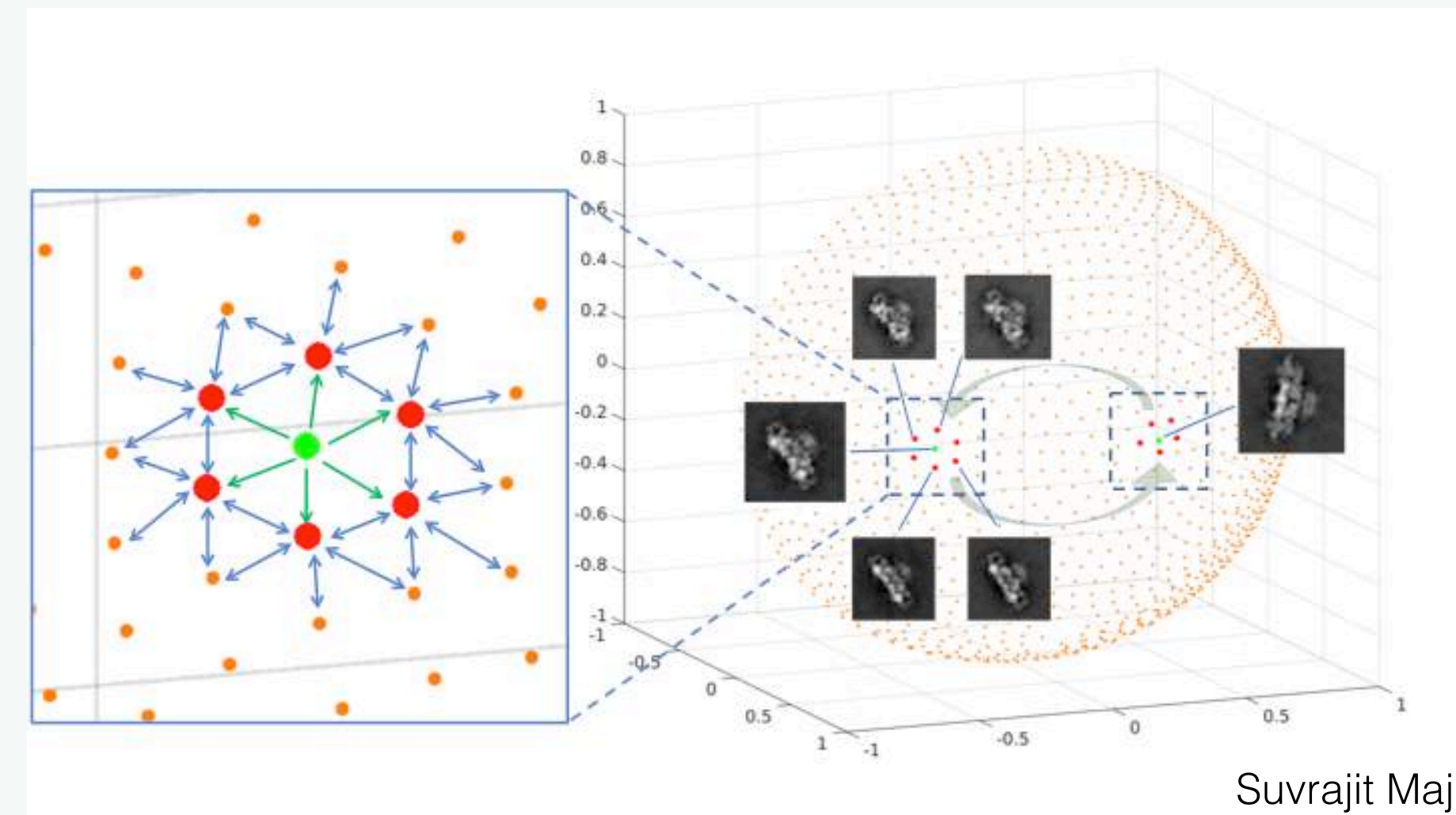
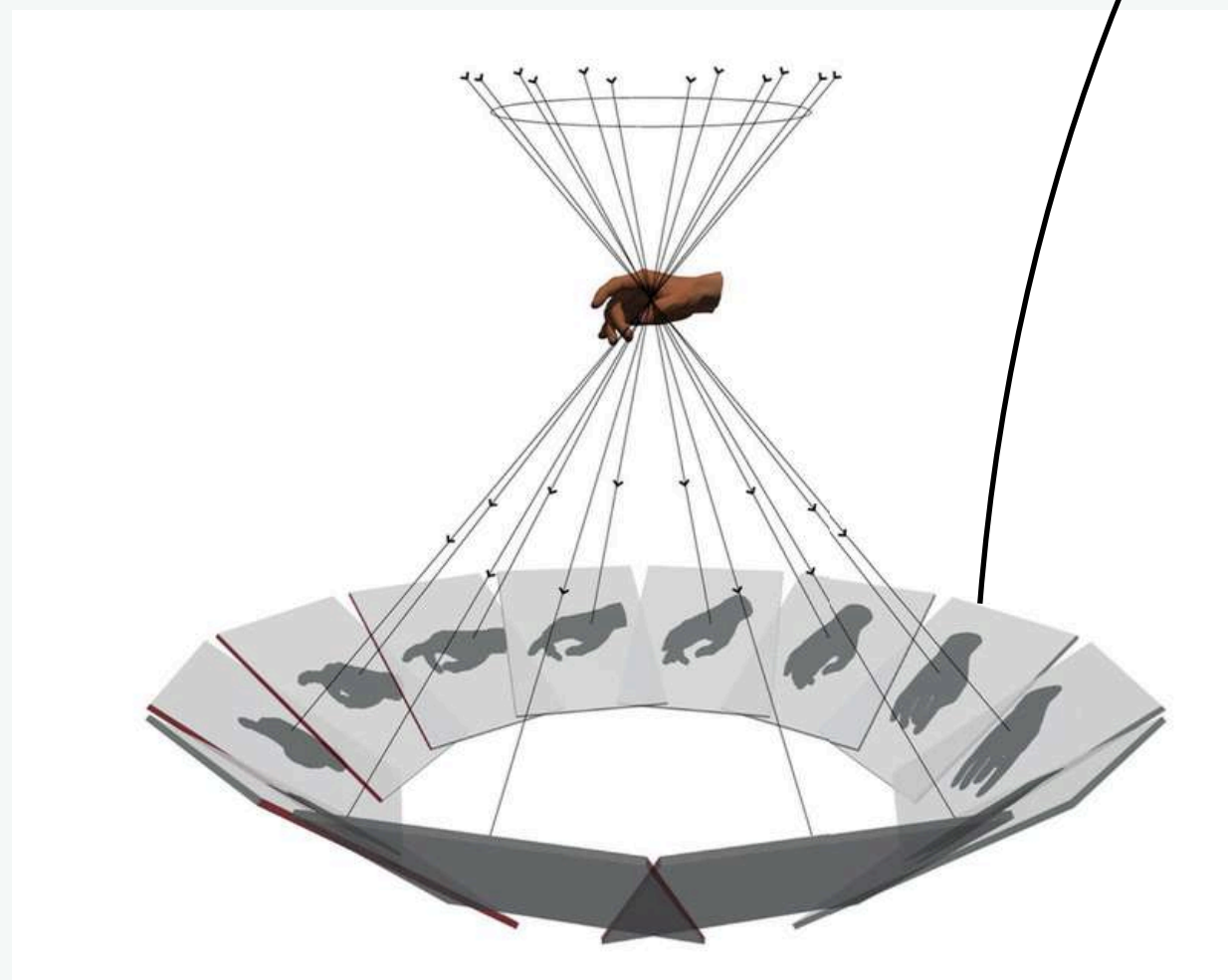


Dashti et al, *Nat Comm* (2020).

One proposed solution: The Manifold Embedding Method in cryo-EM from Ourmazd/Frank labs ¹⁰

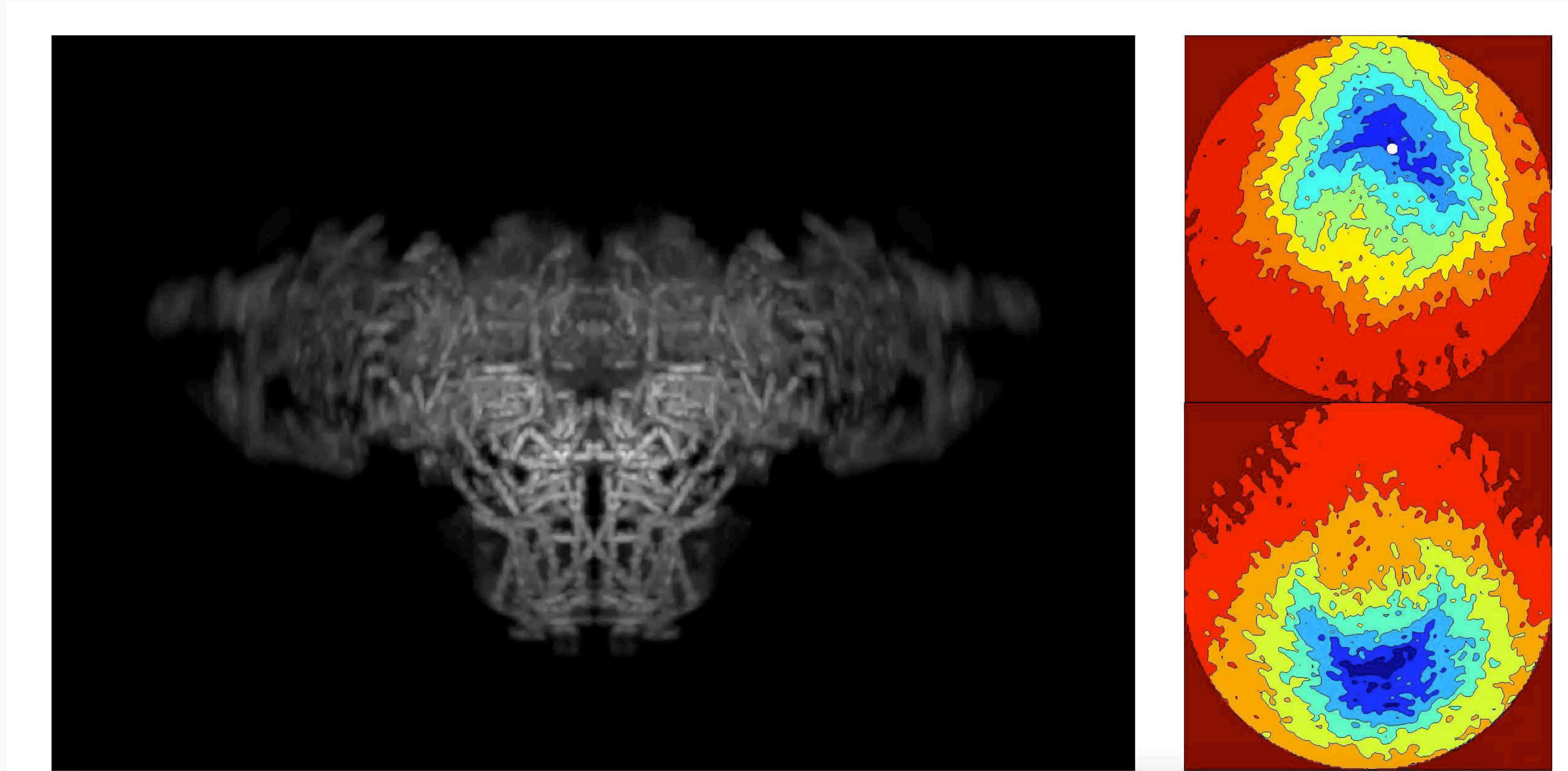
The manifold embedding method uses geometric machine learning to use the information about molecular conformational heterogeneity already available in the particle images.

- 1) Particle images are clustered into 'projection directions'
- 2) The Euclidean distance between each image is calculated
- 3) Spectral decomposition with a diffusion mapping¹ step
- 4) Nonlinear Laplacian spectral analysis (NLSA)² and ordering of images
- 5) Combining the results from individual projection directions into a united depiction of the energy landscape for the whole orientation sphere.³



1. Coifman, R. R., Kevrekidis, I. G., Lafon, S., Maggioni, M. & Nadler, B. Diffusion maps, reduction coordinates, and low dimensional representation of stochastic systems. *Multiscale Model. Simul.* **7**, 842–864 (2008).
2. Giannakis, D. & Majda, A. J. Nonlinear Laplacian spectral analysis for time series with intermittency and low-frequency variability. *Proc. Natl. Acad. Sci.* **109**, 2222–2227 (2012).
3. Maji, S., Liao, H., Dashti, A., Mashayekhi, G., Ourmazd, A., & J. Frank. Propagation of Conformational Coordinates Across Angular Space in Mapping the Continuum of States from Cryo-EM Data by Manifold Embedding. *J. Chem. Inf Model.* (2020).

Manifold Embedding: The Ryanodine Receptor



RELION: Multi-body Refinement



TOOLS AND RESOURCES

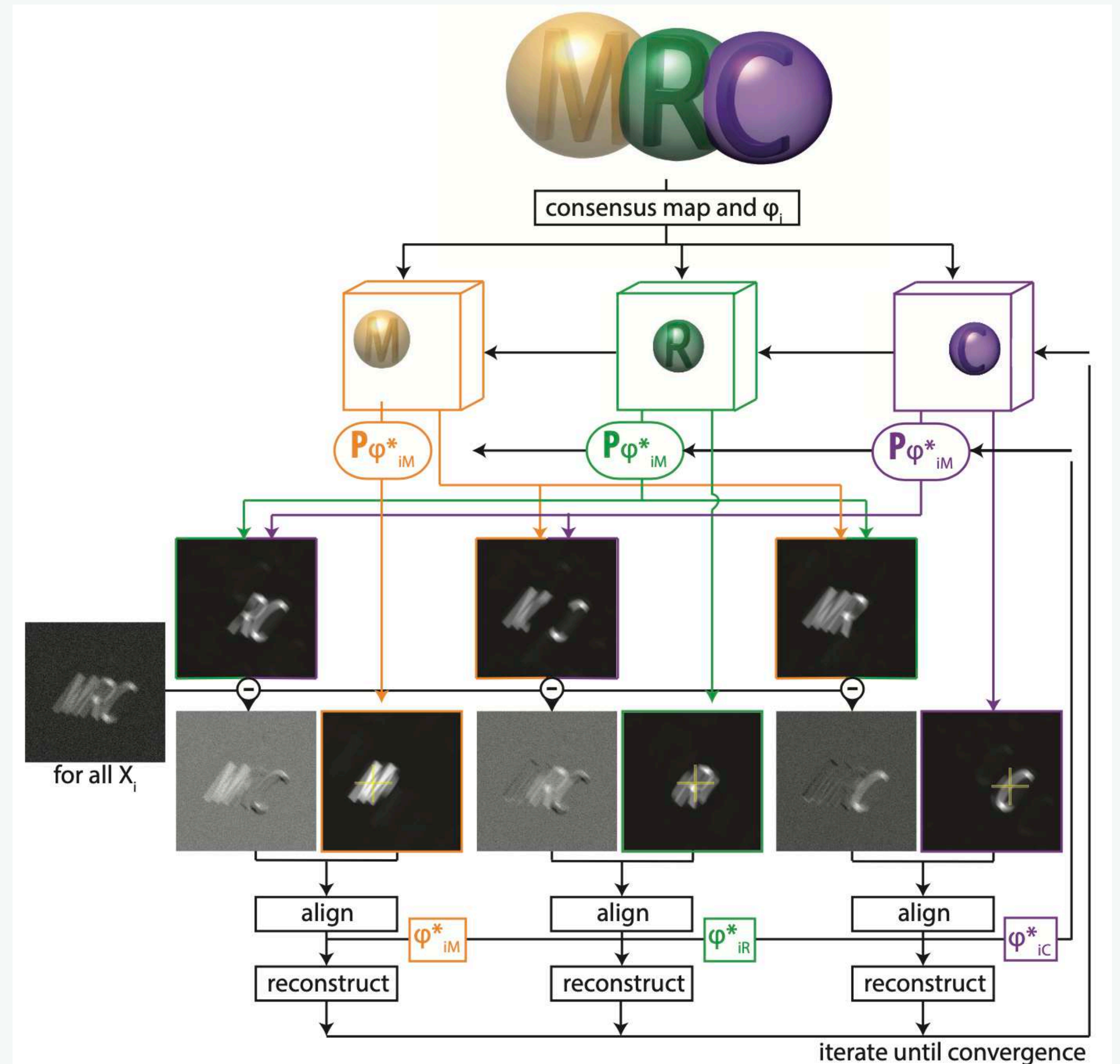
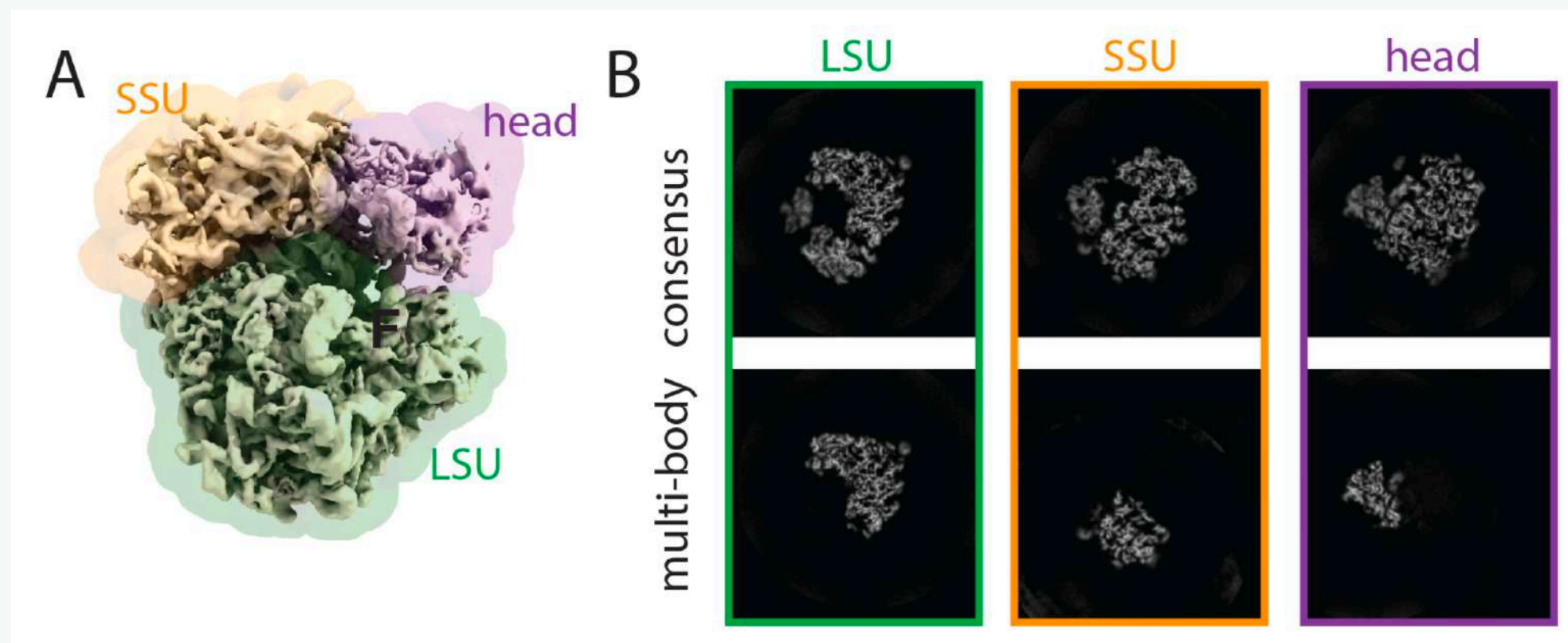


Characterisation of molecular motions in cryo-EM single-particle data by multi-body refinement in RELION

Takanori Nakane¹, Dari Kimanius², Erik Lindahl^{2,3}, Sjors HW Scheres^{1*}

¹MRC Laboratory of Molecular Biology, Cambridge, United Kingdom; ²Department of Biochemistry and Biophysics, Science for Life Laboratory, Stockholm University, Stockholm, Sweden; ³Swedish e-Science Research Center, KTH Royal Institute of Technology, Stockholm, Sweden

Nakane et al. eLife 2018;7:e36861. DOI: <https://doi.org/10.7554/eLife.36861>



cryoDRGN: The first NN approach.

ARTICLES

<https://doi.org/10.1038/s41592-020-01049-4>

nature | methods

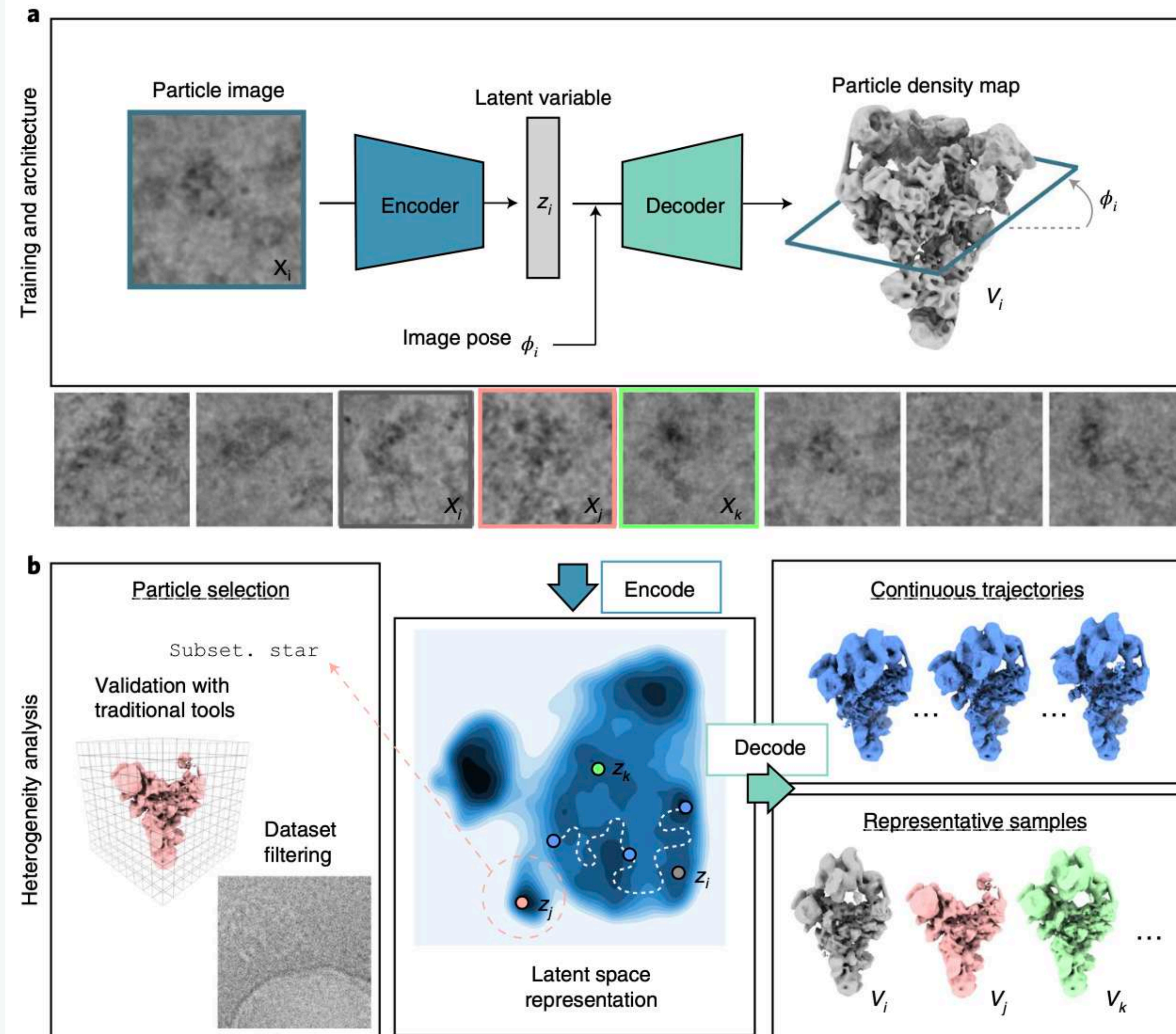
Check for updates

CryoDRGN: reconstruction of heterogeneous cryo-EM structures using neural networks

Ellen D. Zhong^{1,2}, Tristan Bepler^{1,2}, Bonnie Berger^{2,3} and Joseph H. Davis^{1,4}

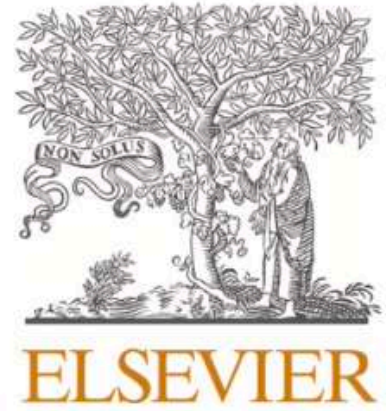
Cryo-electron microscopy (cryo-EM) single-particle analysis has proven powerful in determining the structures of rigid macromolecules. However, many imaged protein complexes exhibit conformational and compositional heterogeneity that poses a major challenge to existing three-dimensional reconstruction methods. Here, we present cryoDRGN, an algorithm that leverages the representation power of deep neural networks to directly reconstruct continuous distributions of 3D density maps and map per-particle heterogeneity of single-particle cryo-EM datasets. Using cryoDRGN, we uncovered residual heterogeneity in high-resolution datasets of the 80S ribosome and the RAG complex, revealed a new structural state of the assembling 50S ribosome, and visualized large-scale continuous motions of a spliceosome complex. CryoDRGN contains interactive tools to visualize a dataset's distribution of per-particle variability, generate density maps for exploratory analysis, extract particle subsets for use with other tools and generate trajectories to visualize molecular motions. CryoDRGN is open-source software freely available at <http://cryodrgn.csail.mit.edu>.

NATURE METHODS | VOL 18 | FEBRUARY 2021 | 176-185 | www.nature.com/naturemethods



3DVA in cryoSPARC

Journal of Structural Biology 213 (2021) 107702



Contents lists available at [ScienceDirect](https://www.sciencedirect.com)

Journal of Structural Biology

journal homepage: www.elsevier.com/locate/yjsbi

3D variability analysis: Resolving continuous flexibility and discrete heterogeneity from single particle cryo-EM

Ali Punjani^{a,b,c,*}, David J. Fleet^{a,b,*}

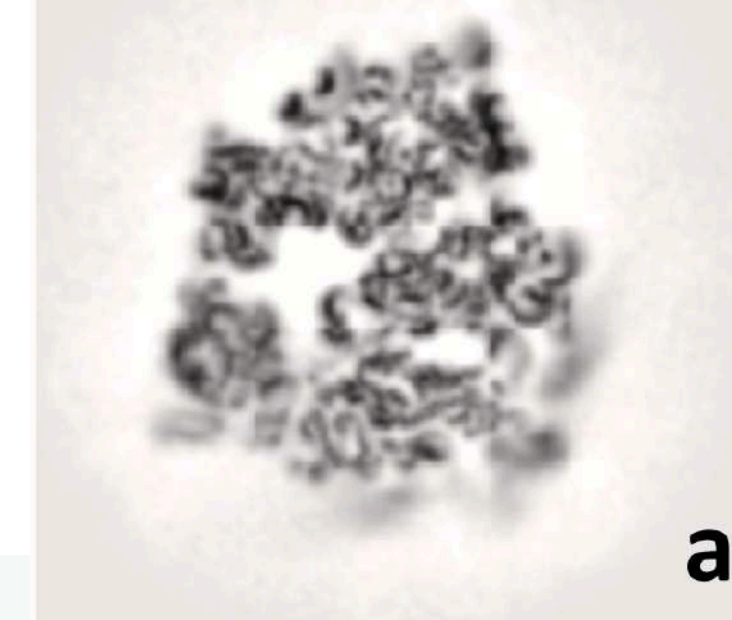
^a Department of Computer Sciences, University of Toronto M5S 3G4, Canada

^b Vector Institute, 710-661 University Ave., Toronto M5G 1M1, Canada

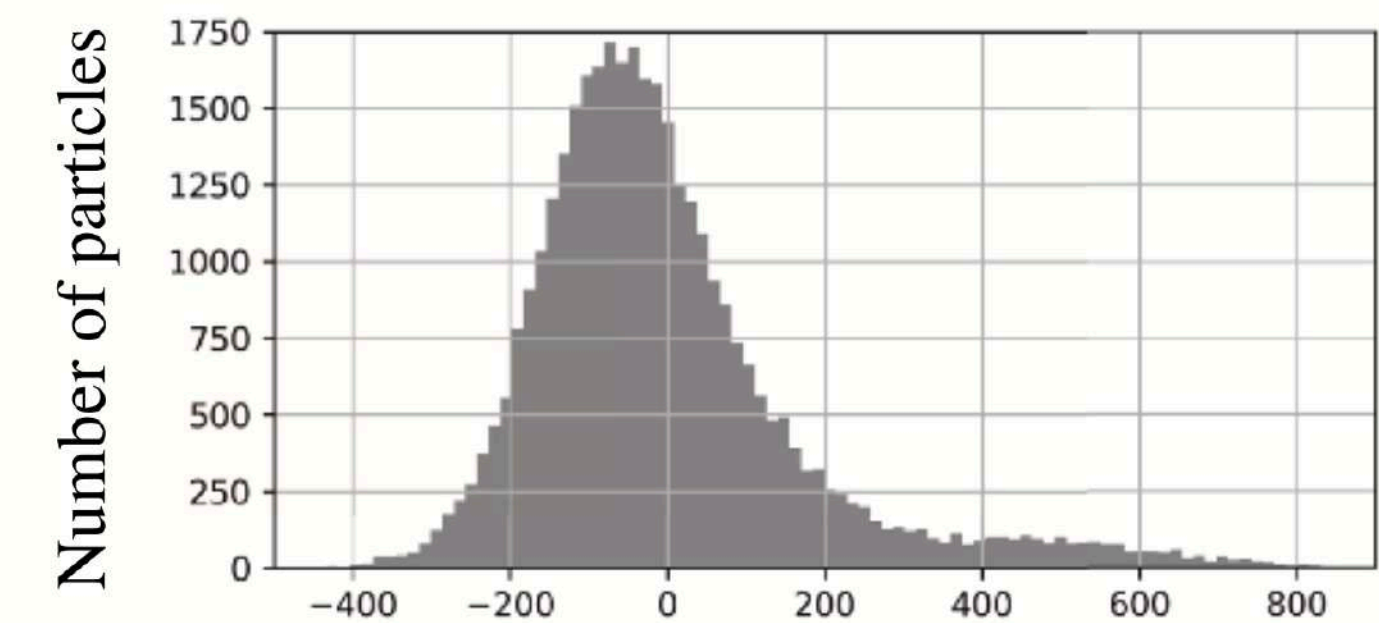
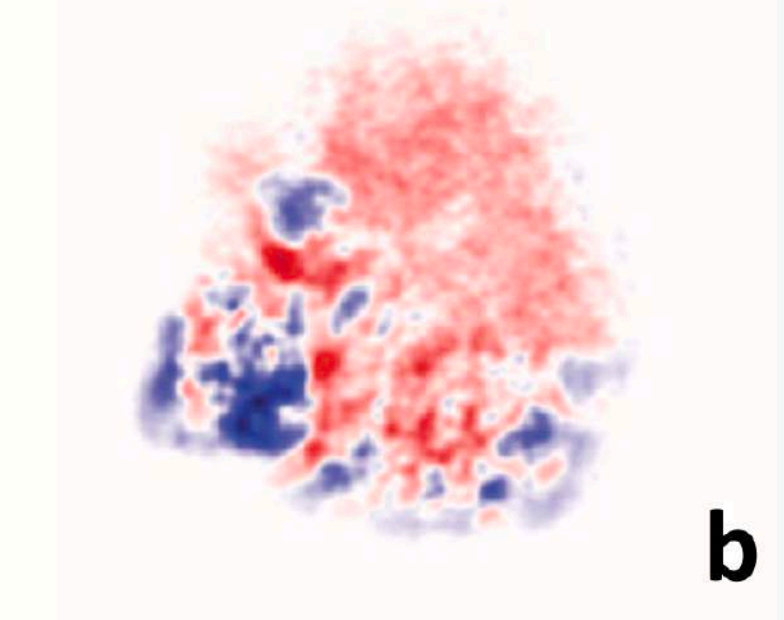
^c Structura Biotechnology Inc., 129-100 College Ave., Toronto M5G 1L5, Canada

- Very straightforward to use.
- Limited to linear motions.

Base 3D Density



Variability component



c Latent coordinate along variability component

Gaussian Mixture Model method - EMAN2

ARTICLES

<https://doi.org/10.1038/s41592-021-01220-5>

nature | methods

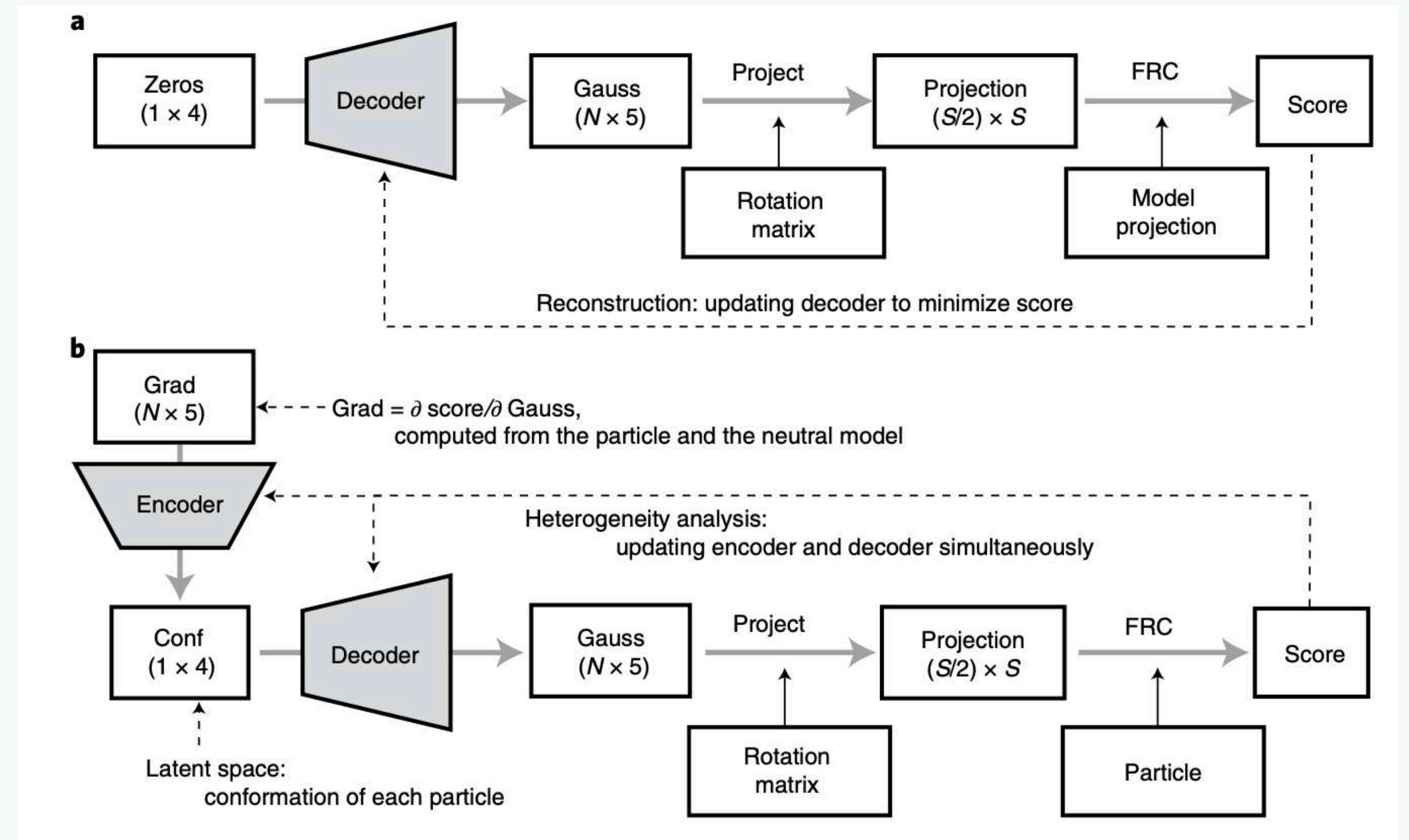
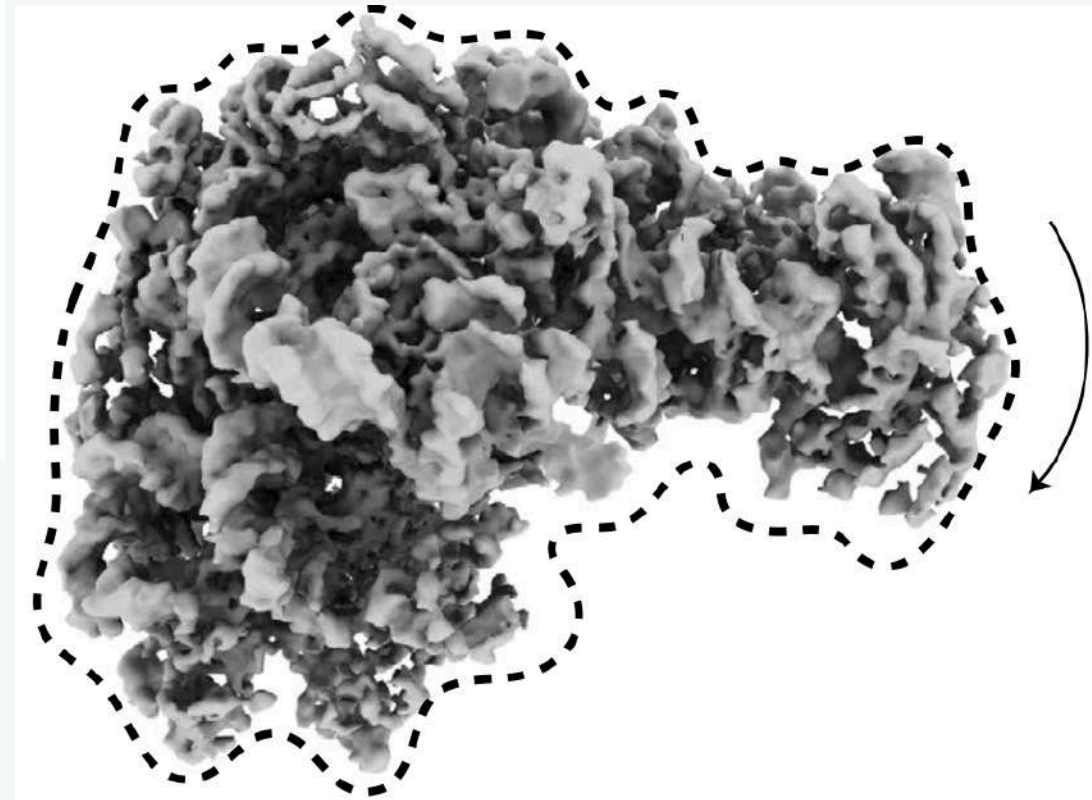
Check for updates

Deep learning-based mixed-dimensional Gaussian mixture model for characterizing variability in cryo-EM

Muyuan Chen  and Steven J. Ludtke  

930

NATURE METHODS | VOL 18 | AUGUST 2021 | 930-936 | www.nature.com/naturemethods



- Incorporating the Gaussian Mixture Model (GMM) gets us toward atomic / molecular information of the dynamics.

CryoSparc's more complex tool for heterogeneity analysis:

nature methods



Article

<https://doi.org/10.1038/s41592-023-01853-8>

3DFlex: determining structure and motion of flexible proteins from cryo-EM

Received: 28 July 2022

Ali Punjani ^{1,2,3} & David J. Fleet ^{1,2,4}

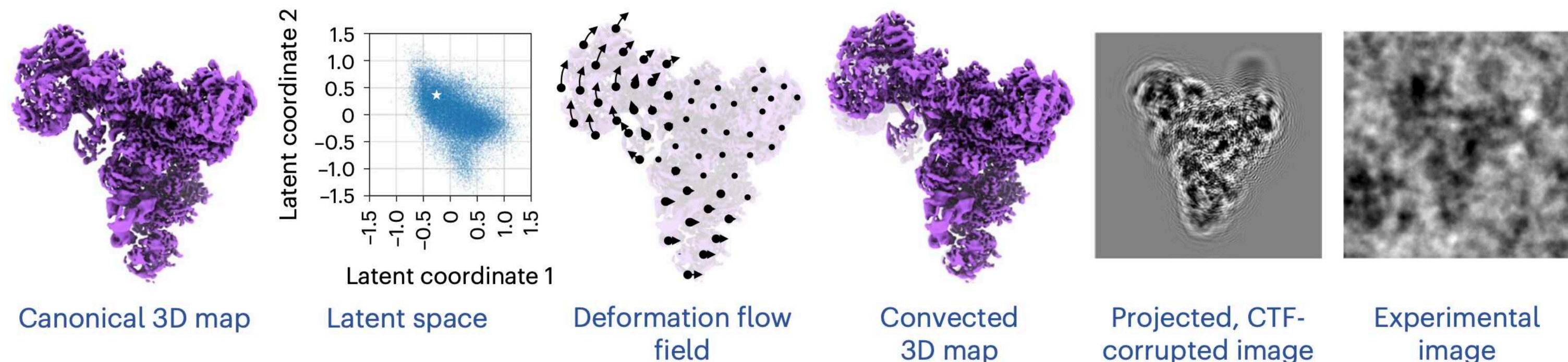
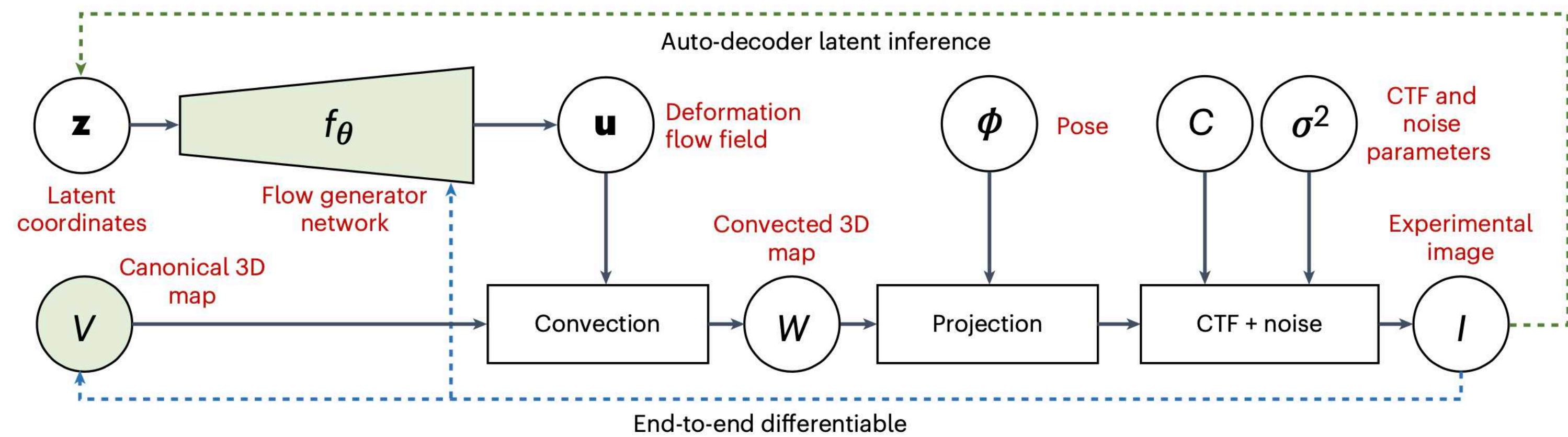
Accepted: 16 March 2023

Published online: 11 May 2023

Check for updates

Modeling flexible macromolecules is one of the single-particle cryogenic-electron microscopy (cryo-EM) to illuminate fundamental questions in structural biology. Here, we present Three-Dimensional Flexible Refinement (3DFlex), a deep learning-based method for cryo-EM reconstruction of flexible macromolecules. 3DFlex is designed to model the flexibility of macromolecules by learning a deformation flow field from a canonical 3D map. The flow field is used to generate a convected 3D map, which is then projected and corrupted with CTF and noise to produce an experimental image. 3DFlex is an end-to-end differentiable pipeline that can be trained on a large dataset of cryo-EM images and used to reconstruct flexible macromolecules from cryo-EM data.

Nature Methods | Volume 20 | June 2023 | 860–870



CryoSparc's more complex tool for heterogeneity analysis:

nature methods

Article

3DFlex: determining the structure of flexible proteins

Received: 28 July 2022

Accepted: 16 March 2023

Published online: 11 May 2023

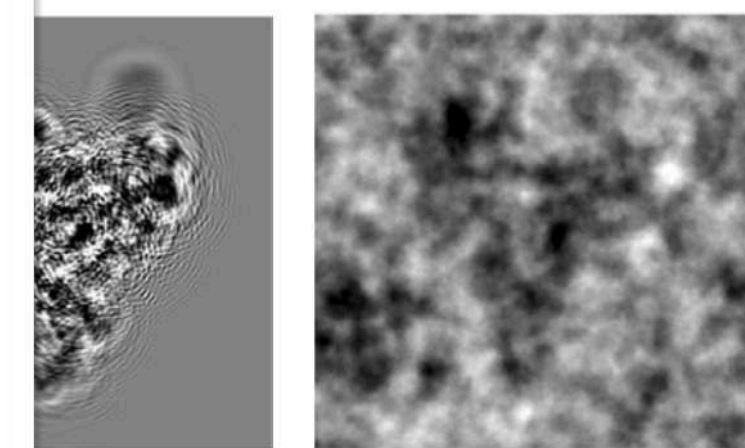
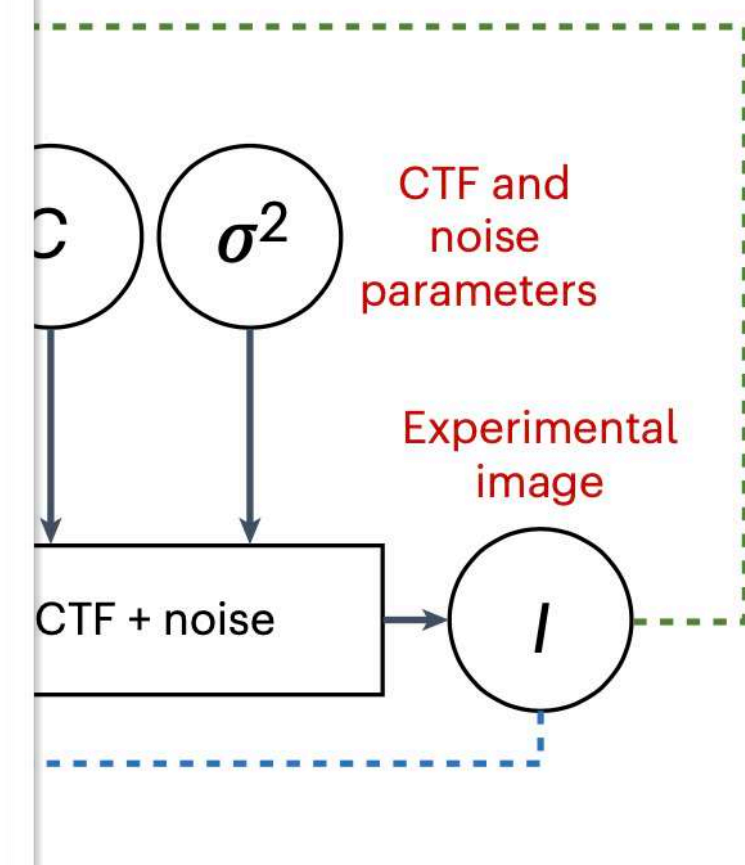
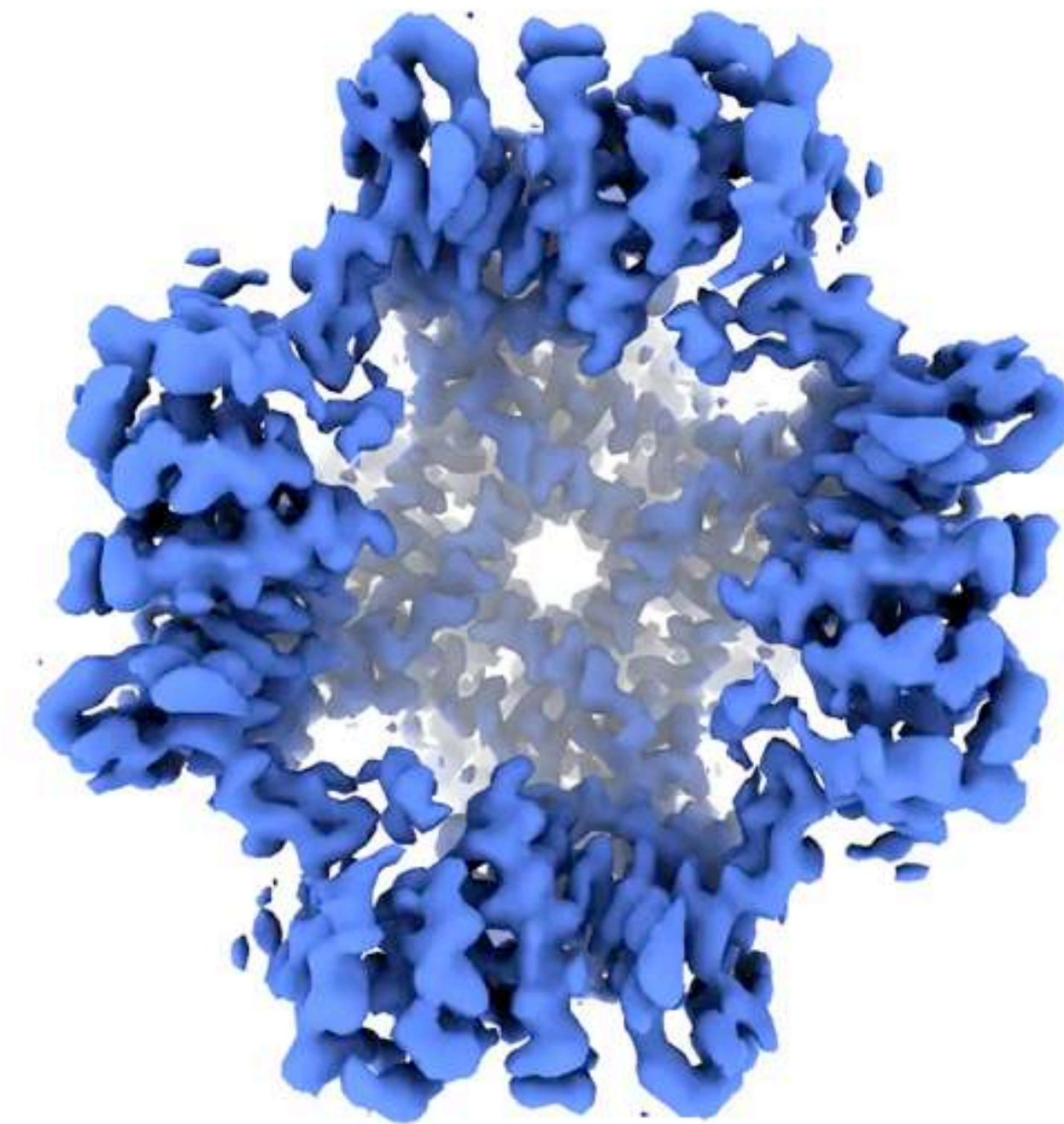
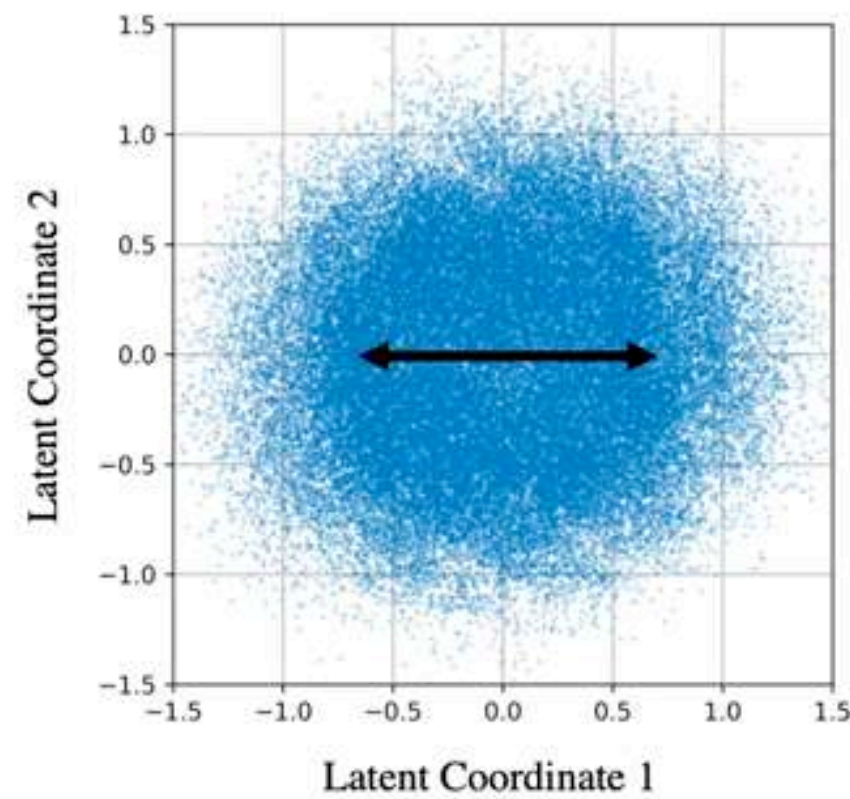
Check for updates

Nature Methods | Volume 20 | June

3D Flexible Refinement

TRPV1 Ion Channel
EMPIAR-10059

Latent coordinate 1



Canonical 3D map

Latent space

Deformation flow field

Convected 3D map

Projected, CTF-corrupted image

Experimental image

RELION also has a new ML-based solution for this problem

Sjors Scheres
@SjorsScheres

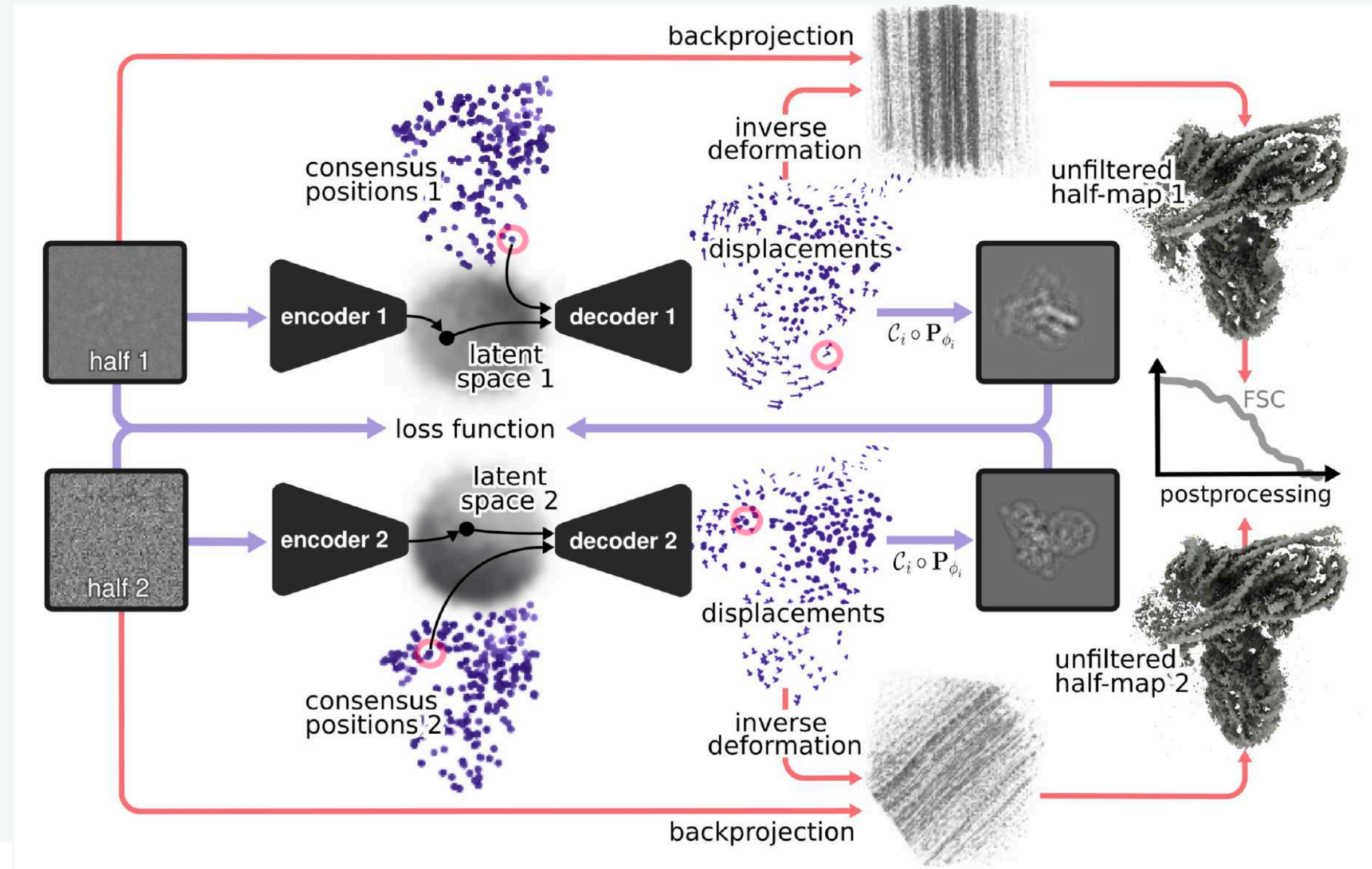
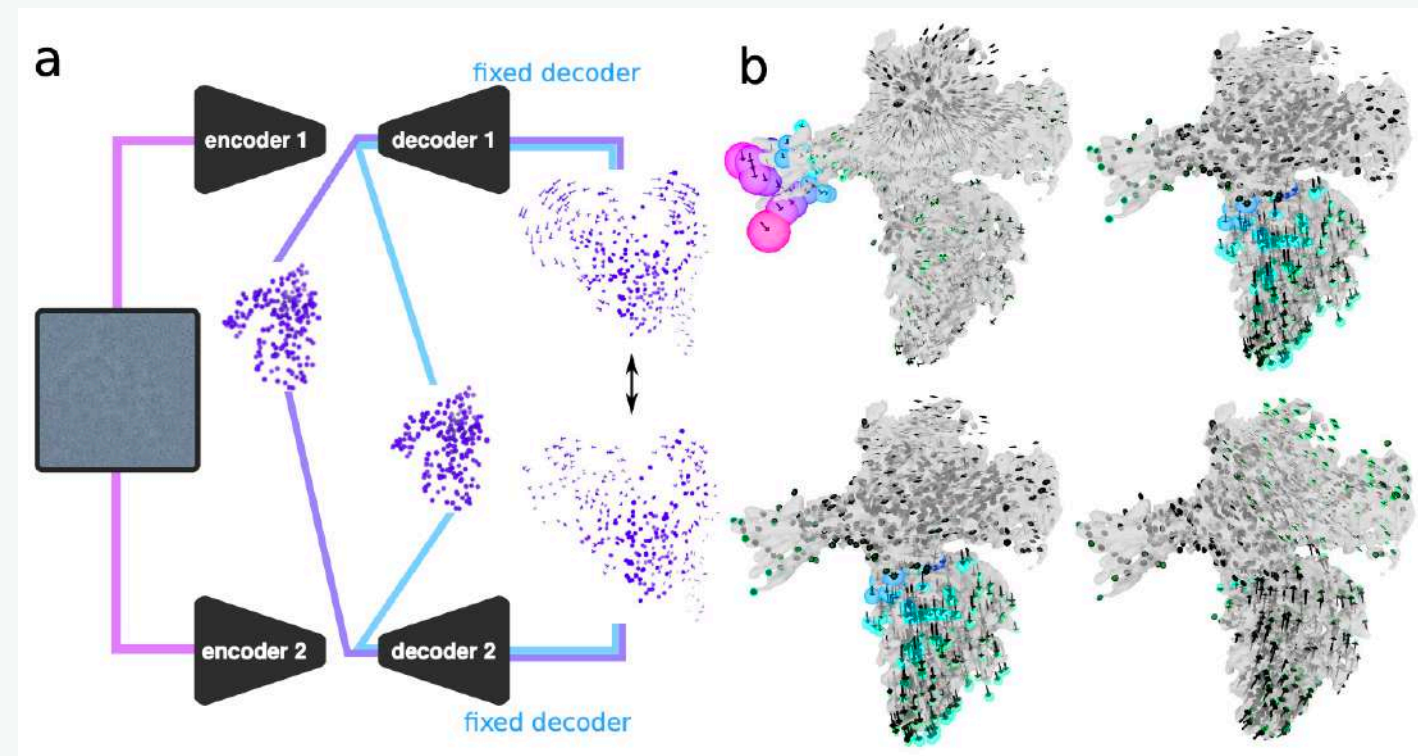
And our new #DynaMight paper is out on @biorxivpreprint! 🥳 This is #RELION5's answer to modelling molecular flexibility by the amazing Johannes Schwab. He also shows that model bias can be nasty when modelling molecular flexibility with many parameters.

bioRxiv
THE PREPRINT SERVER FOR BIOLOGY

biorxiv.org
DynaMight: estimating molecular motions with improved rec
How to deal with continuously flexing molecules is one of the biggest outstanding challenges in single-particle ...

11:29 AM · Oct 19, 2023 · 24.6K Views

Has a method for error estimation of the deformations.



Implemented in RELION-5

Multiple methods available in Scipion

<https://github.com/scipion-em/scipion-em-continuousflex>

nature communications

Nature Communications | (2023)14:154



Article

<https://doi.org/10.1038/s41467-023-35791-y>

Estimating conformational landscapes from Cryo-EM particles by 3D Zernike polynomials

Received: 1 June 2022

Accepted: 29 December 2022

Published online: 11 January 2023

Check for updates

D. Herreros¹ , R. R. Lederman², J. M. Krieger¹, A. Jiménez-Moreno¹, M. Martínez¹, D. Myška³, D. Strelak^{1,4}, J. Filipovic³, C. O. S. Sorzano^{1,5} & J. M. Carazo^{1,5}

The new developments in Cryo-EM Single Particle Analysis are helping us to understand how the macromolecular structure and function meet to drive

jmb
Journal of Molecular Biology



Volume 435, Issue 9, 1 May 2023, 167951

Research Article

MDSPACE: Extracting Continuous Conformational Landscapes from Cryo-EM Single Particle Datasets Using 3D-to-2D Flexible Fitting based on Molecular Dynamics Simulation

Rémi Vuillemot^{1,6}, Alex Mirzaei¹, Mohamad Harastani¹, Ilyes Hamitouche¹, Léo Fréchin², Bruno P. Klaholz², Osamu Miyashita³, Florence Tama^{3,4,5}, Isabelle Rouiller⁶, Slavica Jonic¹

TikTok and Cows?

bioRxiv preprint doi: <https://doi.org/10.1101/2023.10.31.564872>; this version posted December 7, 2023. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under a [CC-BY-NC-ND 4.0 International license](#).

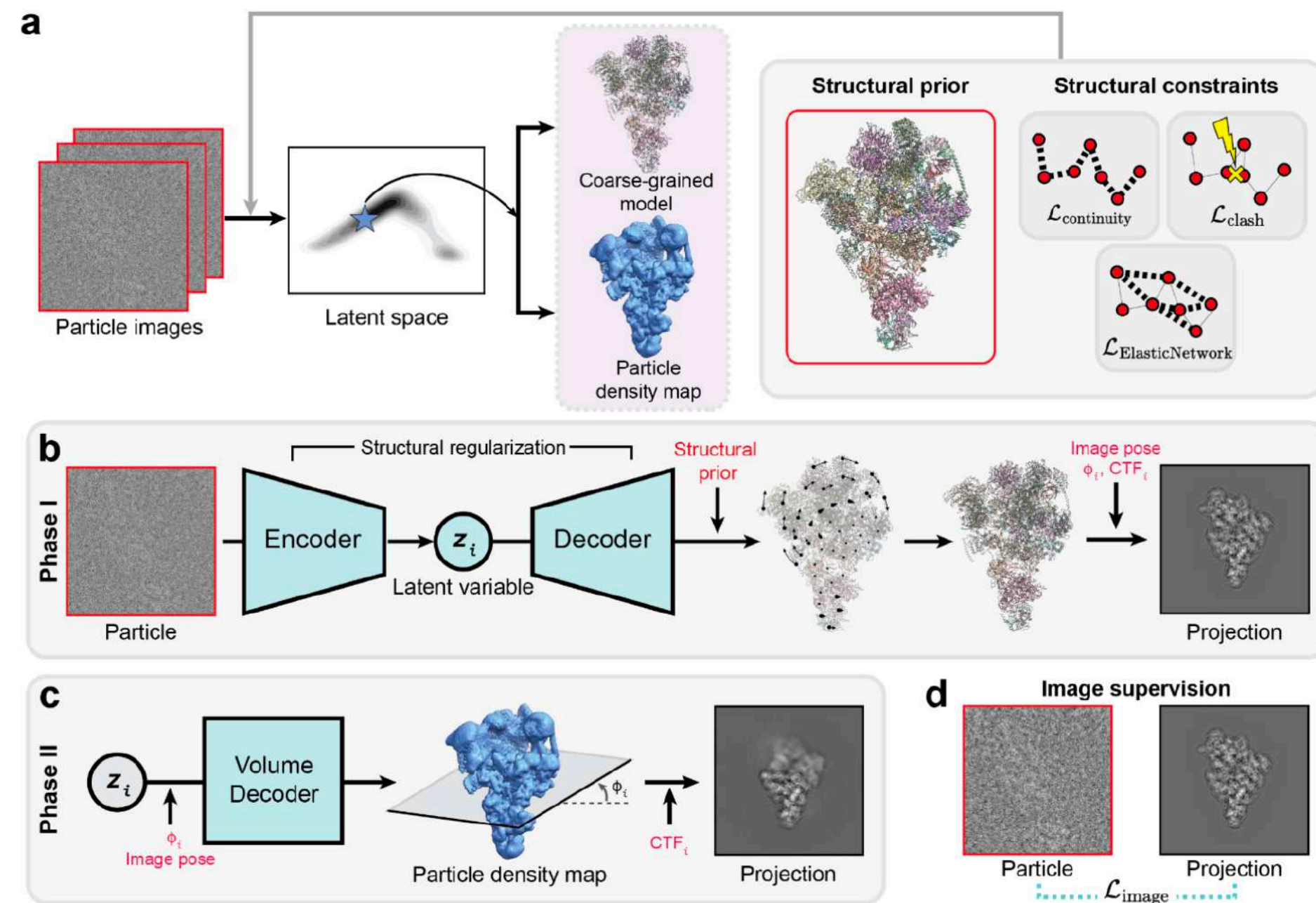
CryoSTAR: Leveraging Structural Prior and Constraints for Cryo-EM Heterogeneous Reconstruction

Yilai Li^{1#}, Yi Zhou^{1#}, Jing Yuan^{1#}, Fei Ye¹, Quanquan Gu^{1*}

¹ByteDance Research

#Contributed Equally

*Correspondence to: quanquan.gu@bytedance.com



CowScape: Quantitative reconstruction of the conformational landscape of biological macromolecules from cryo-EM data

Felix Lambrecht¹, Andreas Kröpelin², Mario Lüttich¹, Michael Habeck^{2,1,*},
David Haselbach^{1,3,*}, Holger Stark^{1,*}

February 20, 2024

¹Max Planck Institute for Multidisciplinary Sciences, 37077 Göttingen, Germany

²Microscopic Image Analysis Group, Jena University Hospital, 07743 Jena, Germany

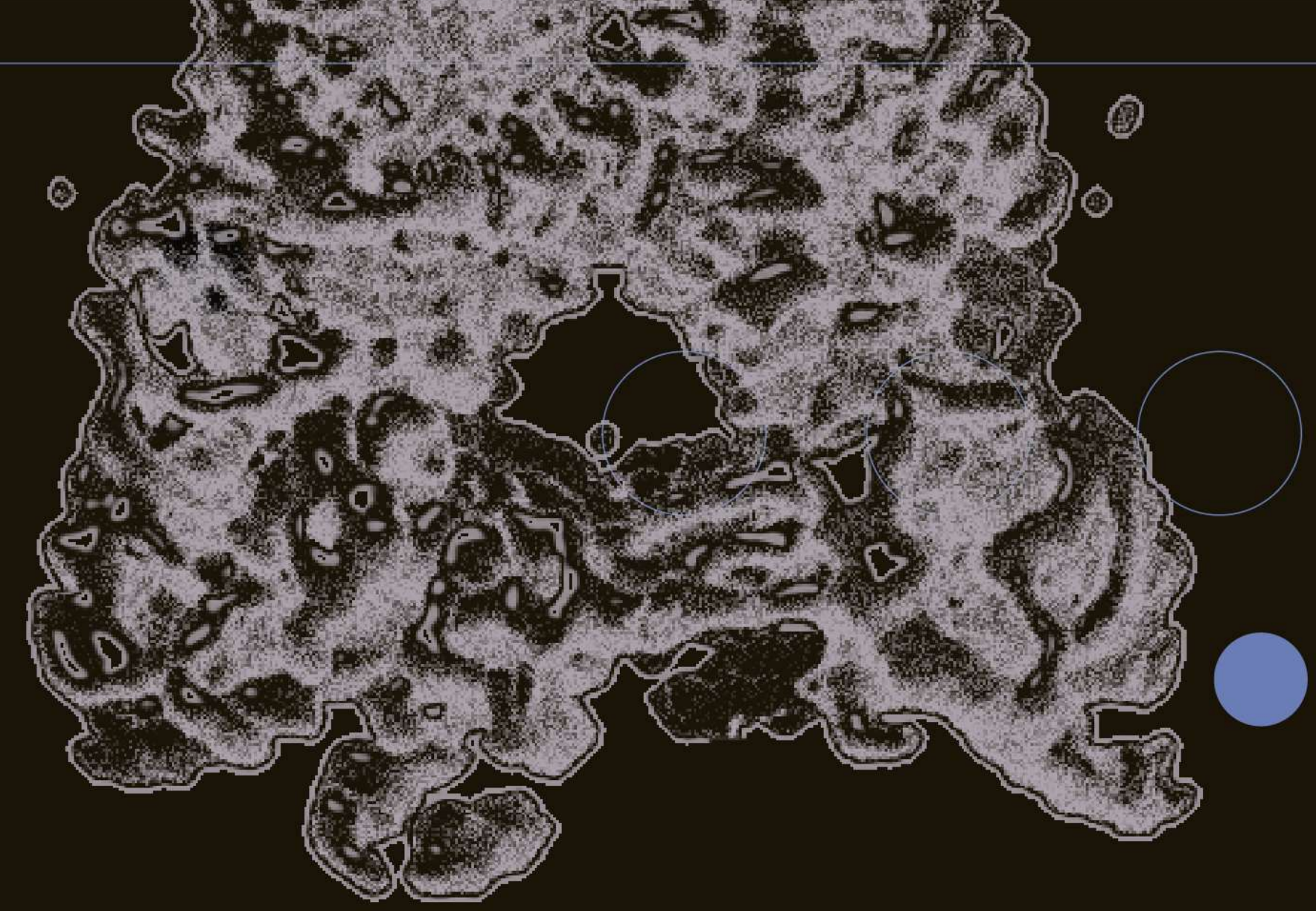
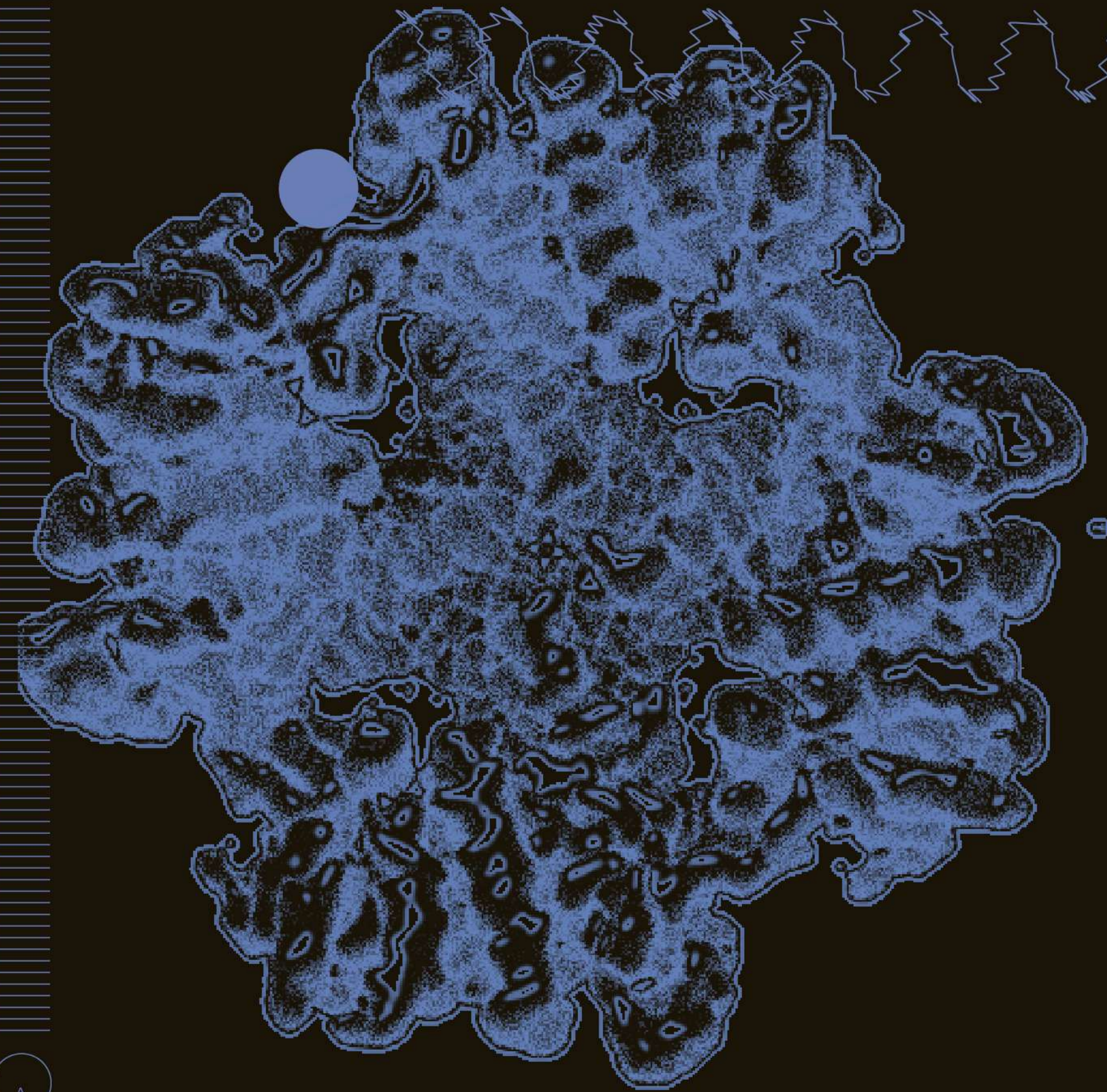
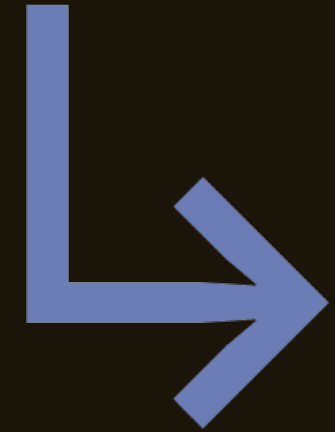
³Institute for Molecular Pathology, Vienna, Austria

*E-Mail: hstark1@gwdg.de; david.haselbach@imp.ac.at; michael.habeck@uni-jena.de

Abstract

Cryo-EM data processing typically focuses on the structure of the main conformational state under investigation and discards images that belong to other states. This approach can reach atomic resolution, but ignores vast amounts of valuable information about the underlying conformational ensemble and its dynamics. CowScape analyzes an entire cryo-EM dataset and thereby obtains a quantitative description of structural variability of macromolecular complexes that represents the biochemically relevant conformational space. By combining extensive image classification with principal component analysis (PCA) of the classified 3D volumes and kernel density estimation, CowScape can be used as a quantitative tool to analyze this variability. PCA projects all 3D structures along the major modes spanning a low-dimensional space that captures a large portion of structural variability. The number of particle images in a given state can be used to calculate an energy landscape based on kernel density estimation and Boltzmann

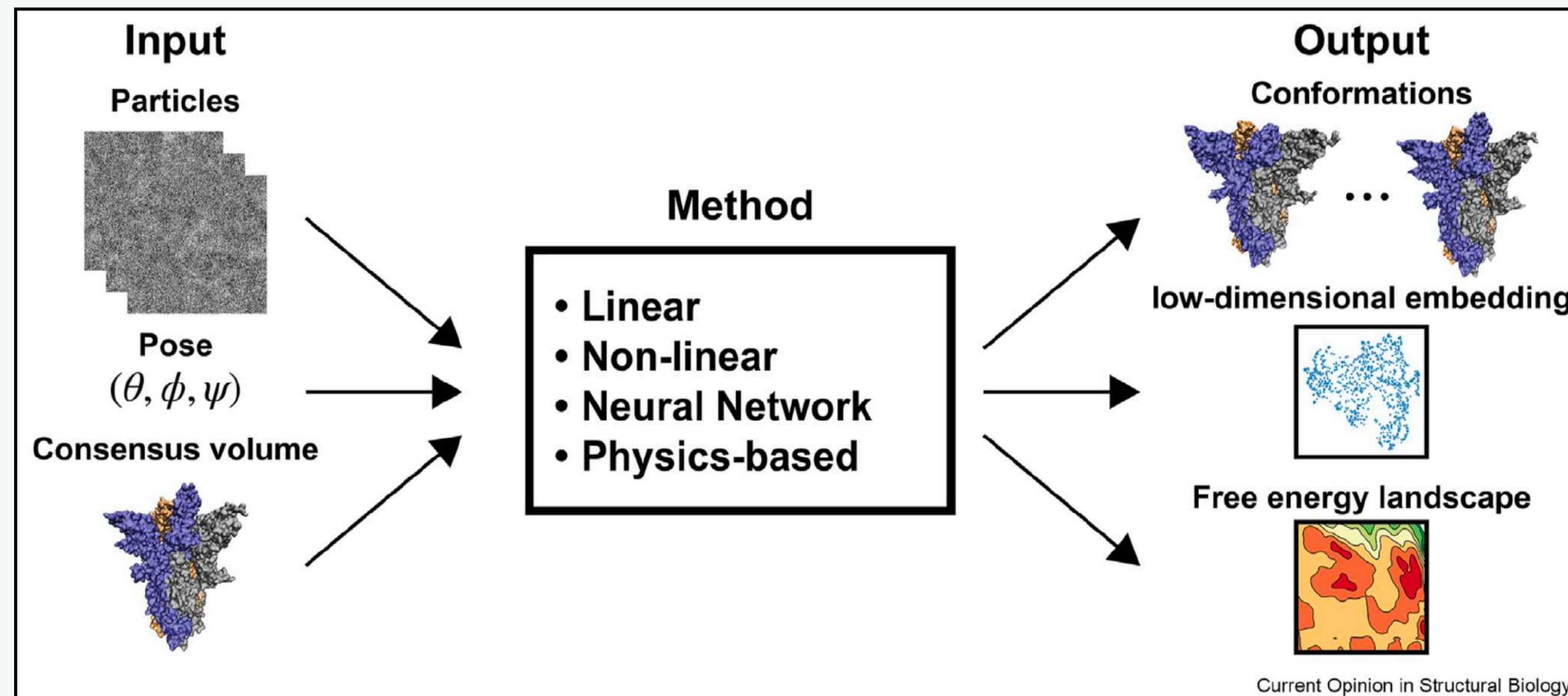
arXiv:2402.11589v1 [q-bio.BM] 18 Feb 2024



Practical considerations when using continuous heterogeneity methods.



Most methods take in particle stacks after 3D Refinement from a previous tool.



- Many methods are already integrated into other tools that already do 3D Refinement (cryoSPARC/RELION/EMAN2) so this is straightforward
- cryoDRGN has a friendly easily installable Python pipeline and plays a little nicer with cryoSPARC inputs

Procedures to jointly optimize pose and conformation are still being improved.



This ICCV paper is the Open Access version, provided by the Computer Vision Foundation.

Except for this watermark, it is identical to the accepted version; ICCV 2021
the final published version of the proceedings is available on IEEE Xplore.

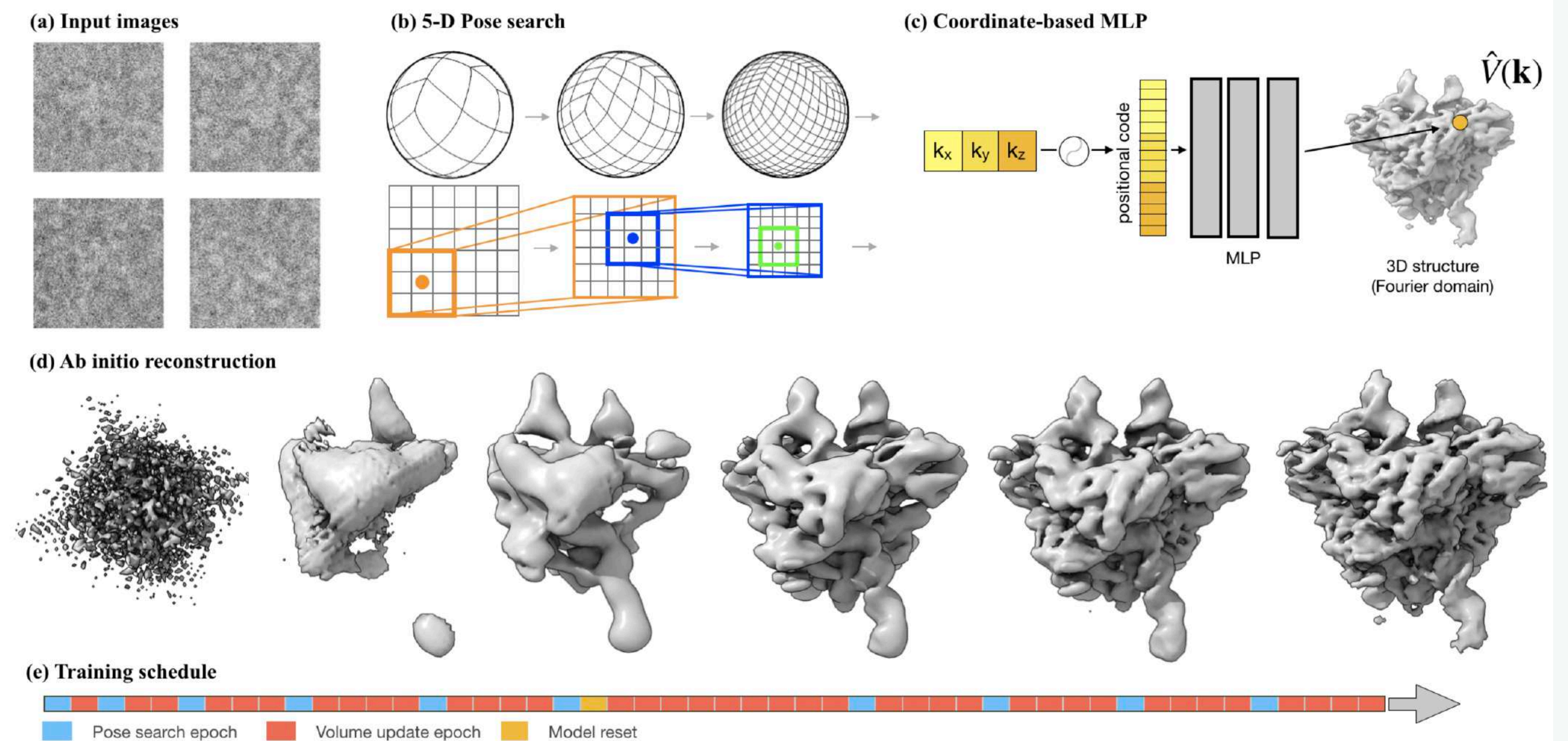
CryoDRGN2: *Ab initio* neural reconstruction of 3D protein structures from real cryo-EM images

Ellen D. Zhong
MIT
zhonge@mit.edu

Adam Lerer
Facebook AI
alerer@fb.com

Joseph H. Davis
MIT
jhdavis@mit.edu

Bonnie Berger
MIT
bab@mit.edu



Procedures to jointly optimize pose and conformation are still being improved.

IEEE TRANSACTIONS ON COMPUTATIONAL IMAGING, VOL. 7, 2021

759

CryoGAN: A New Reconstruction Paradigm for Single-Particle Cryo-EM Via Deep Adversarial Learning

Harshit Gupta , Michael T. McCann , *Member, IEEE*, Laurène Donati , and Michael Unser , *Fellow, IEEE*

Amortized Inference for Heterogeneous Reconstruction in Cryo-EM


36th Conference on Neural Information Processing Systems (NeurIPS 2022).

Axel Levy 
Stanford University

Gordon Wetzstein 
Stanford University

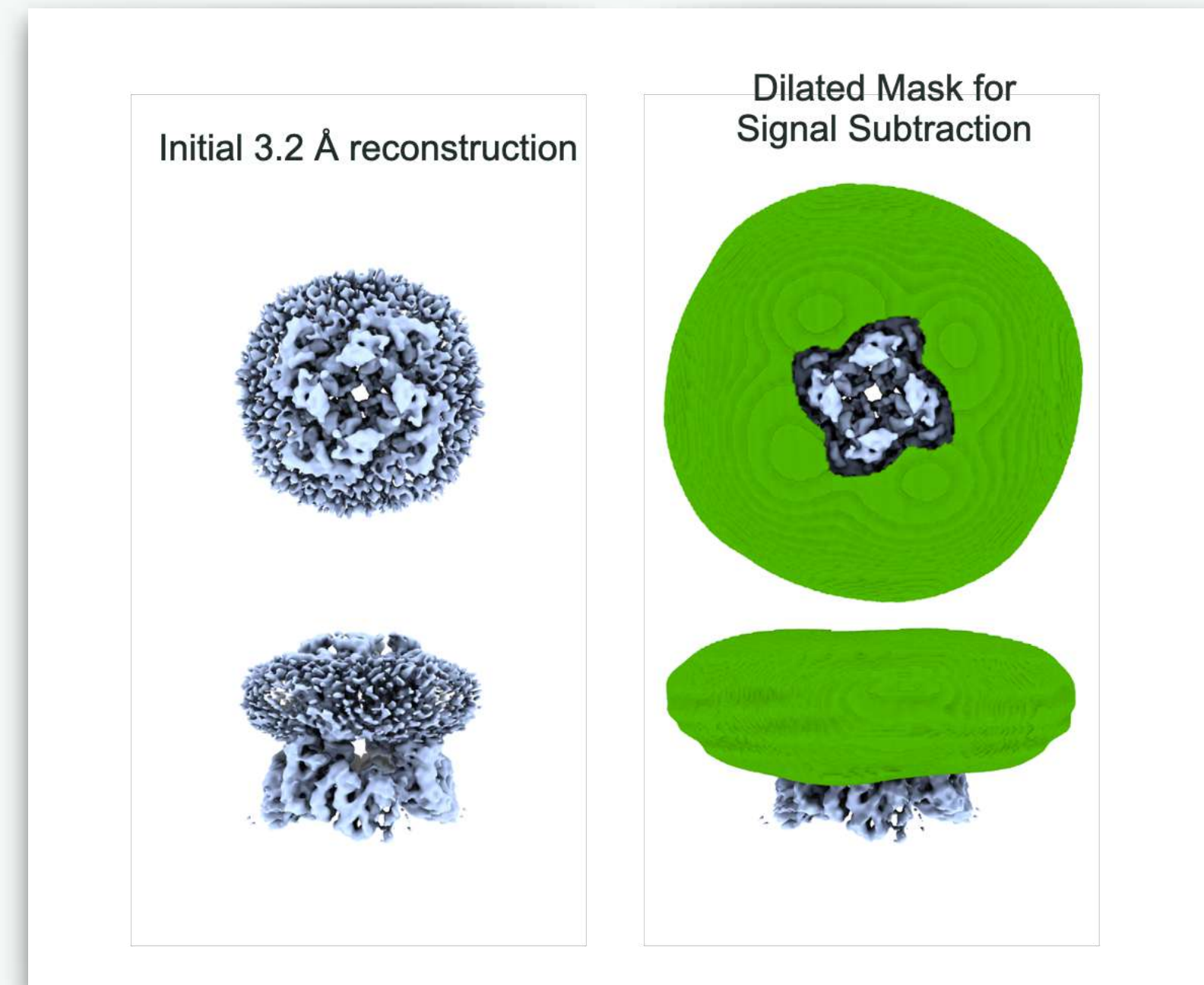
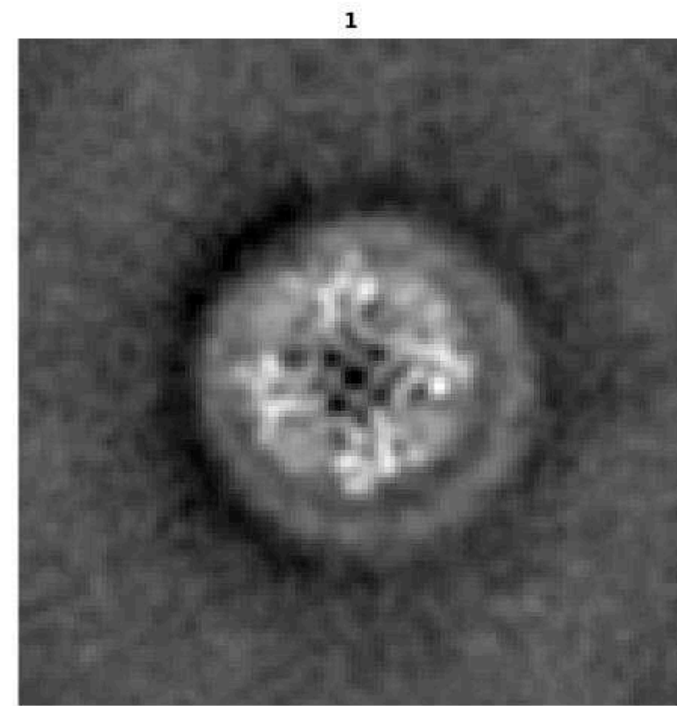
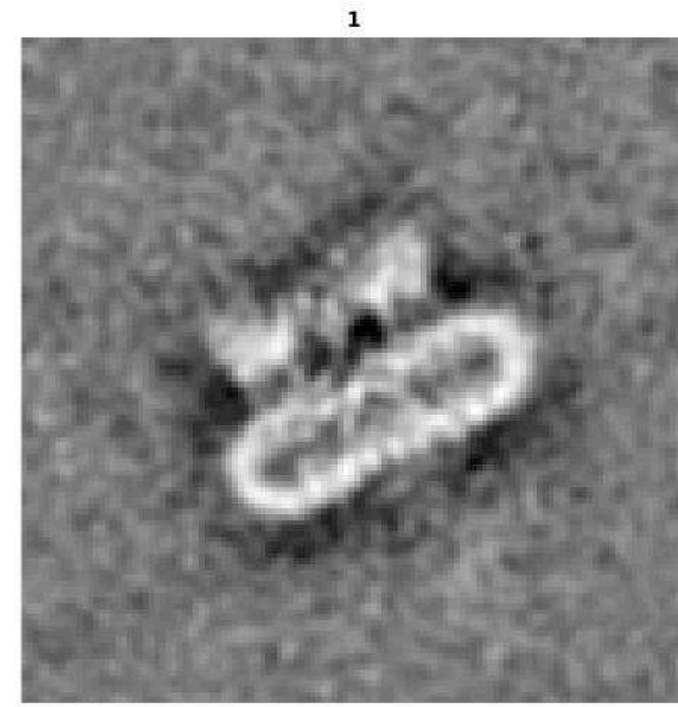
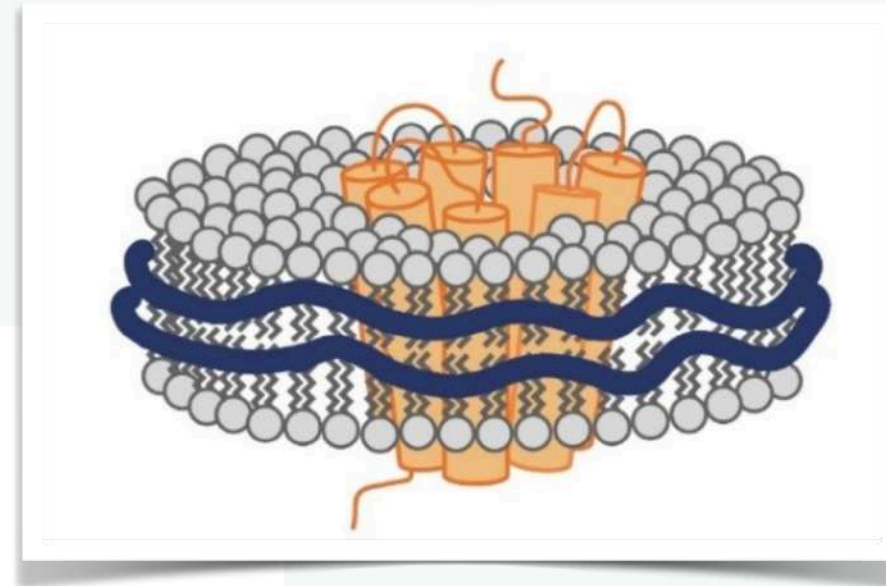
Julien Martel 
Stanford University

Frédéric Poitevin 
SLAC National Accelerator Laboratory

Ellen D. Zhong *
Princeton University

Nanodiscs can also be a challenge for continuous heterogeneity methods.

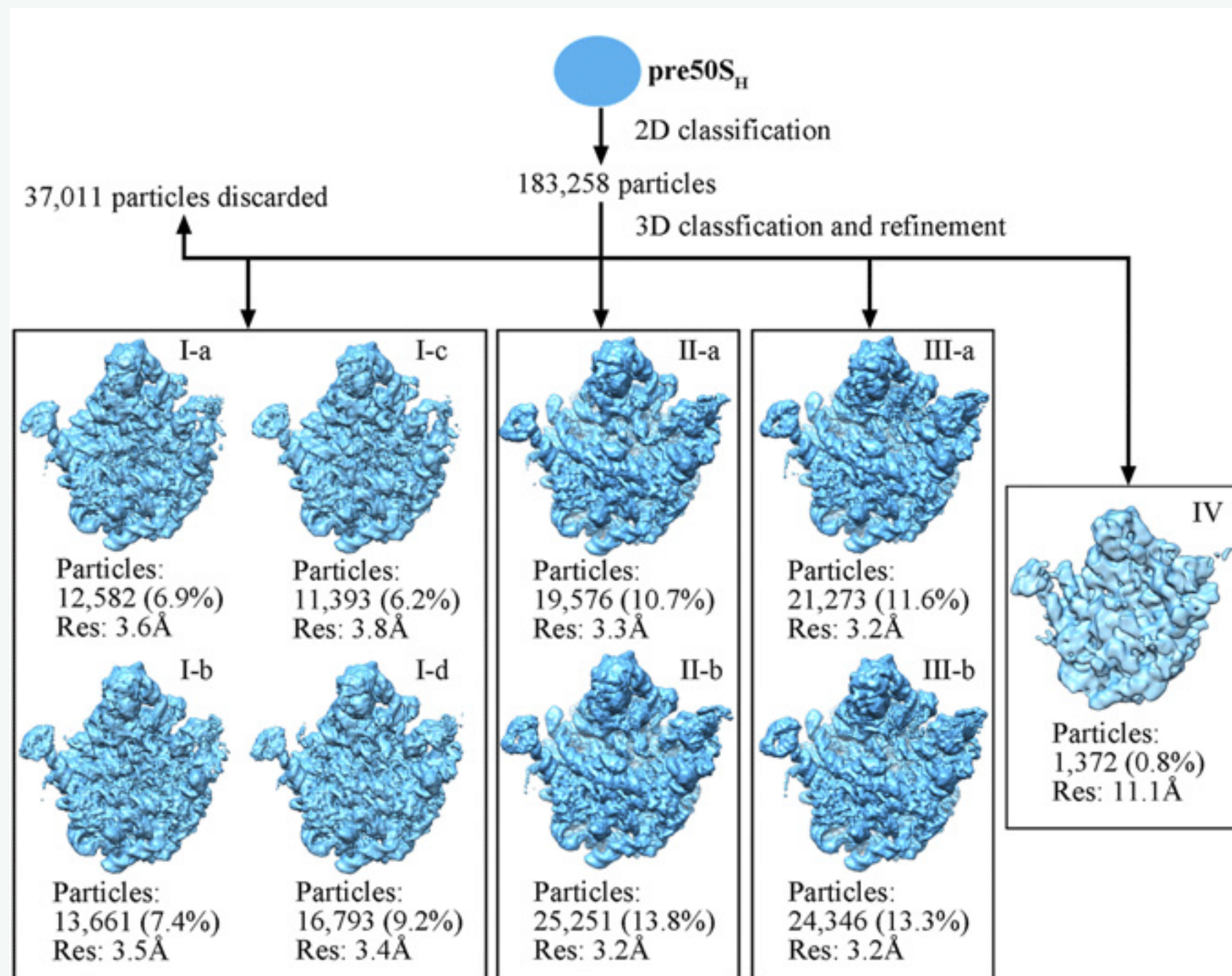
rTRPV1 + DkTx/RtX
in nanodiscs
(EMPIAR 10059)



Analysis with ManifoldEM Matlab Code

Using all vs. a subset of your data.

- Getting rid of junk is important.
- Making sure there isn't residual compositional heterogeneity is important
- Intuitively you wouldn't want particles too tightly classified in a single conformational state, but this is a little tricky to assess



nature communications Nature Communications | (2023)14:7822

Article <https://doi.org/10.1038/s41467-023-43555-x>

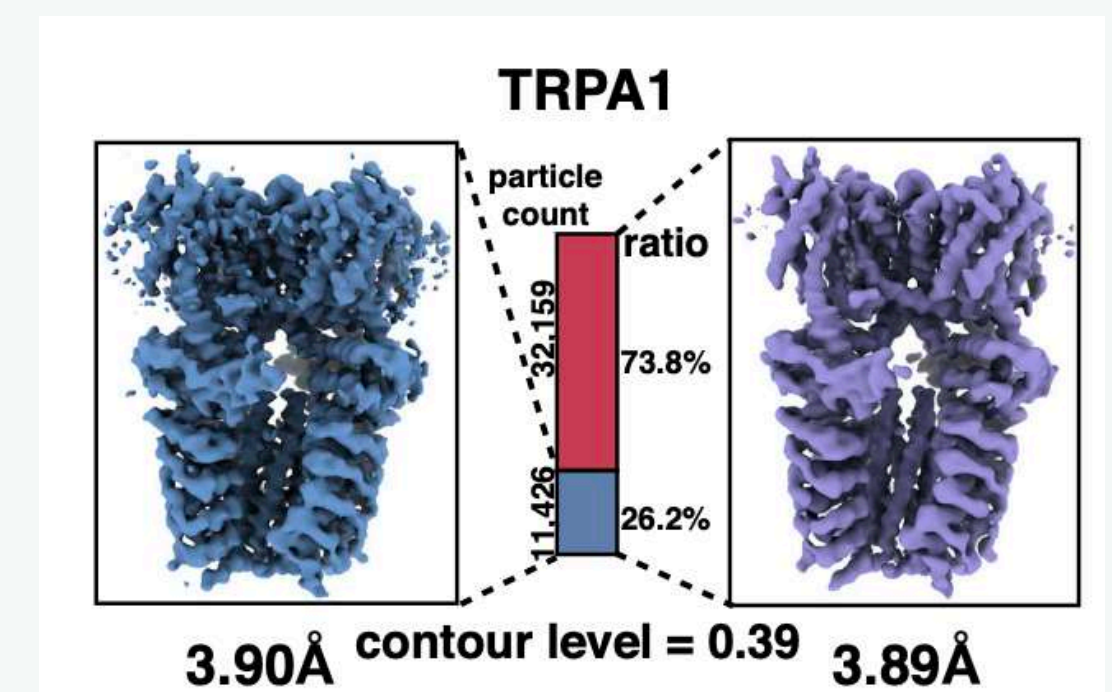
A minority of final stacks yields superior amplitude in single-particle cryo-EM

Received: 19 May 2023 Accepted: 13 November 2023 Published online: 10 December 2023

Jiaying Zhu^{1,10}, Qi Zhang^{2,3,4,5,10}, Hui Zhang⁶, Zuoqiang Shi^{1,7}, Mingxu Hu^{2,3,4,5,8} & Chenglong Bao^{1,7,9}

Cryogenic electron microscopy (cryo-EM) is widely used to determine near-atomic resolution structures of biological macromolecules. Due to the low signal-to-noise ratio, cryo-EM relies on averaging many images. However, a crucial question in the field of cryo-EM remains unanswered: how close can we get to the minimum number of particles required to reach a specific resolution

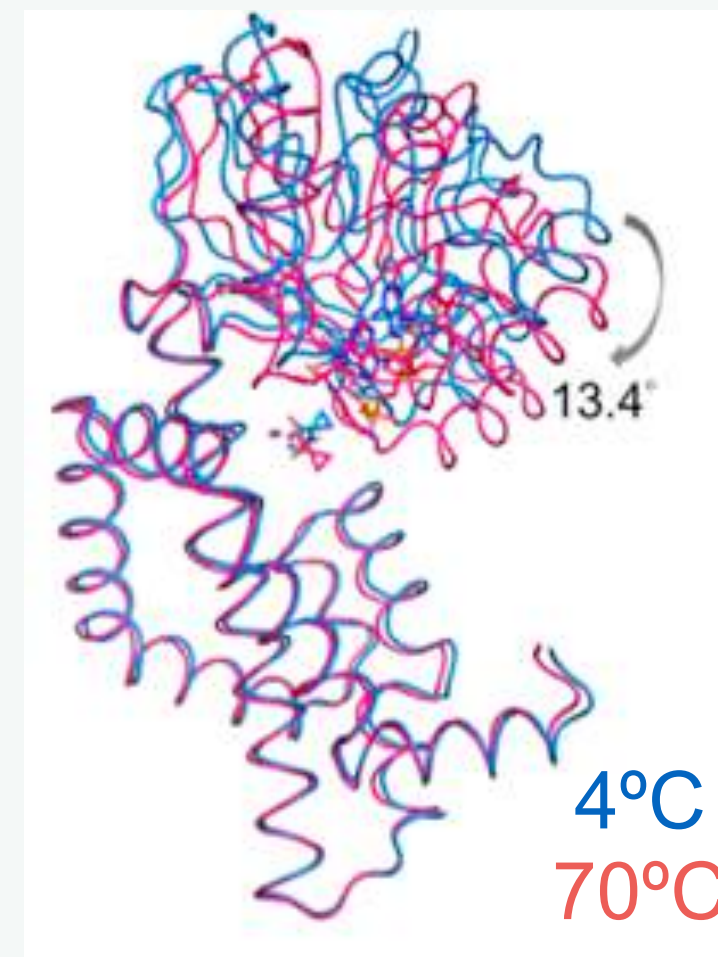
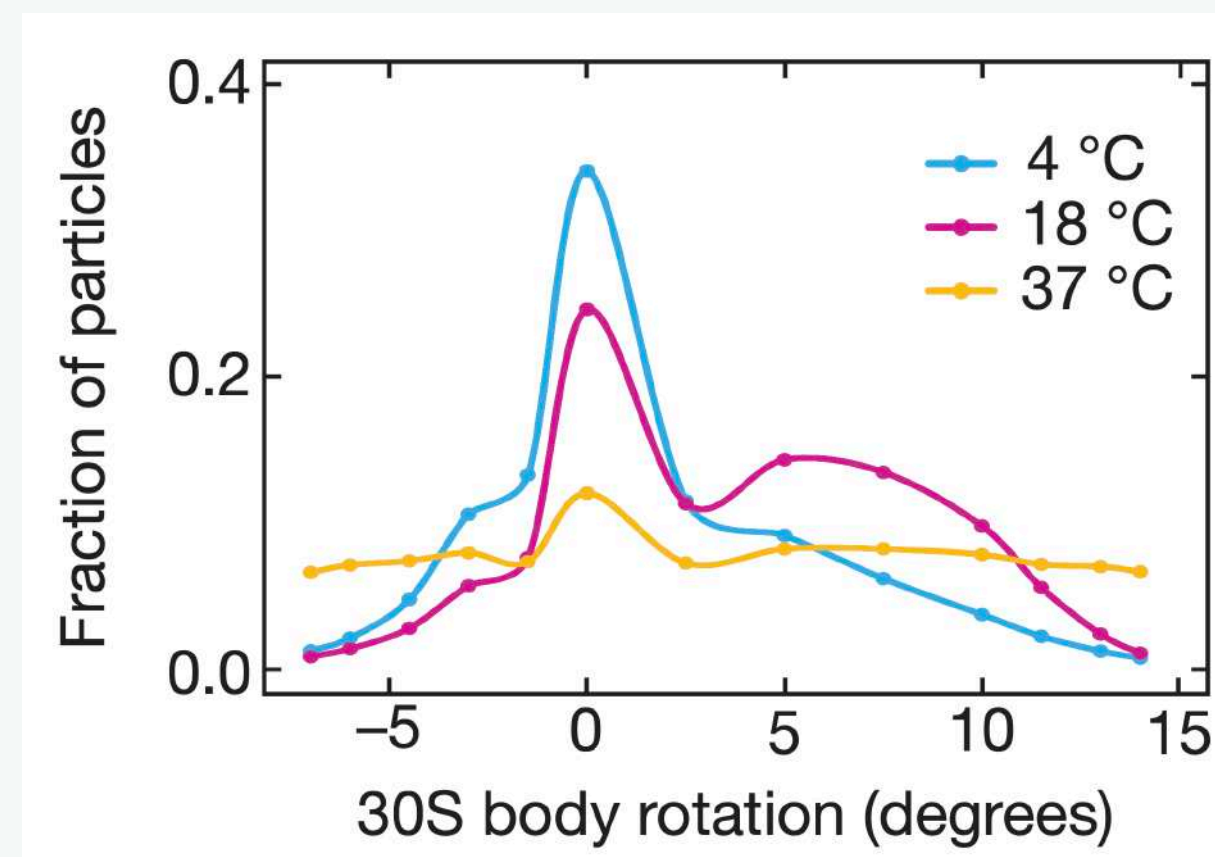
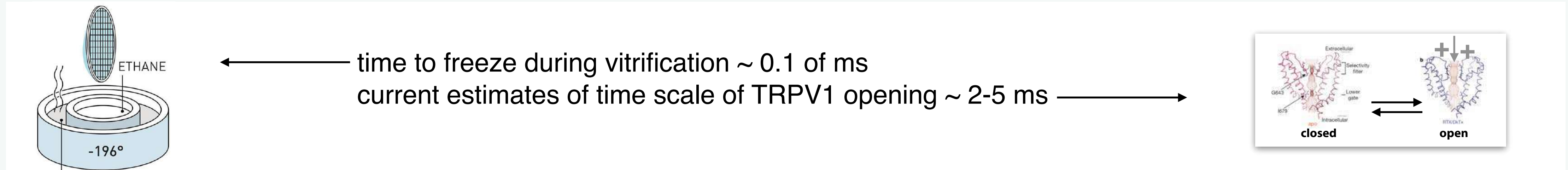
CryoSieve is a new tool that allows you to build a same resolution reconstruction with a fraction of the particles.



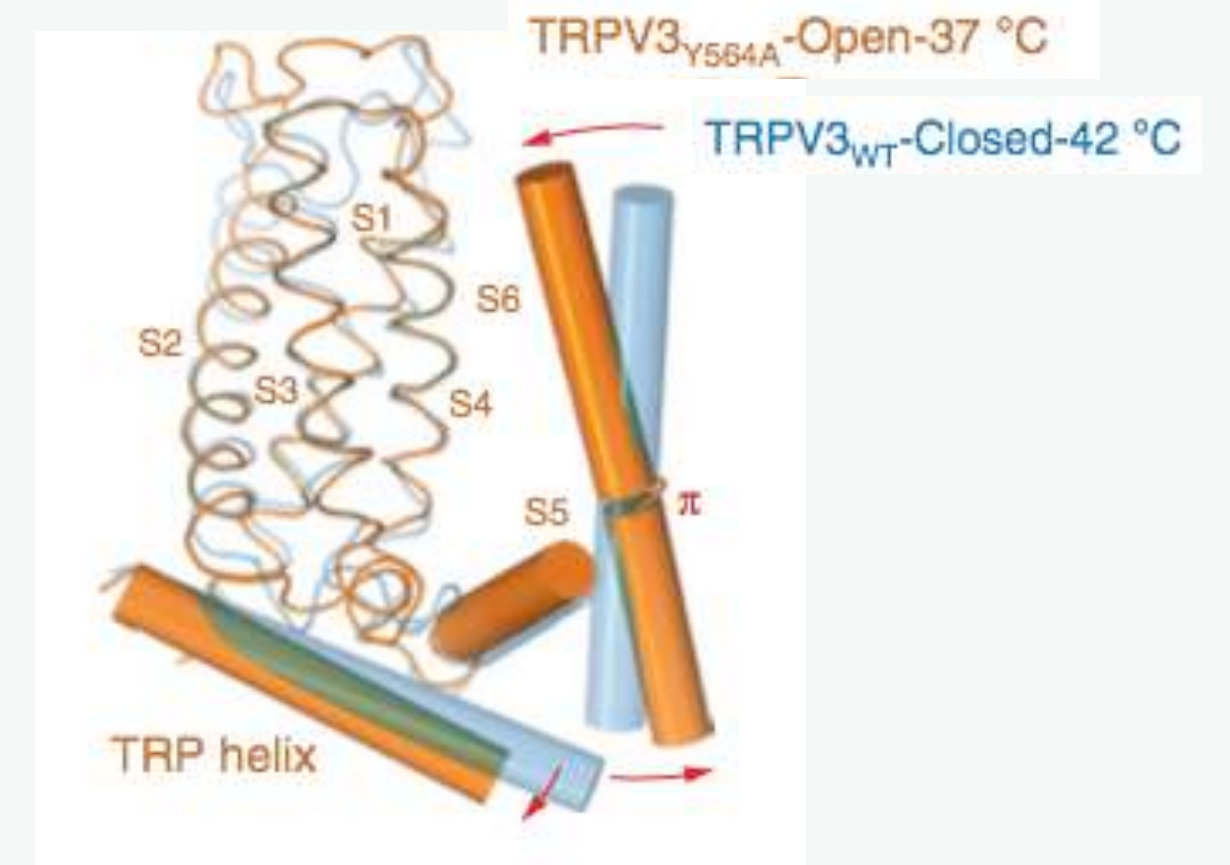


Combining cryo-EM and Molecular Dynamics Simulations

How much to worry about the 'cryo' of cryo-EM?



Chen, Chang, Lin, Huang & Tsai *JACS* (2019).



At 37°C, the majority of the TRPV3_{Y564A} particles were classified into low-resolution reconstructions, likely representing an ensemble of heterogeneous conformations.

Fischer, Konevega, Wintermeyer, Rodnina, & Stark, *Nature* (2010)

Singh... Sobolevsky, *Nature Struc Mol Biol* (2019).

MD Simulations can give some insight...



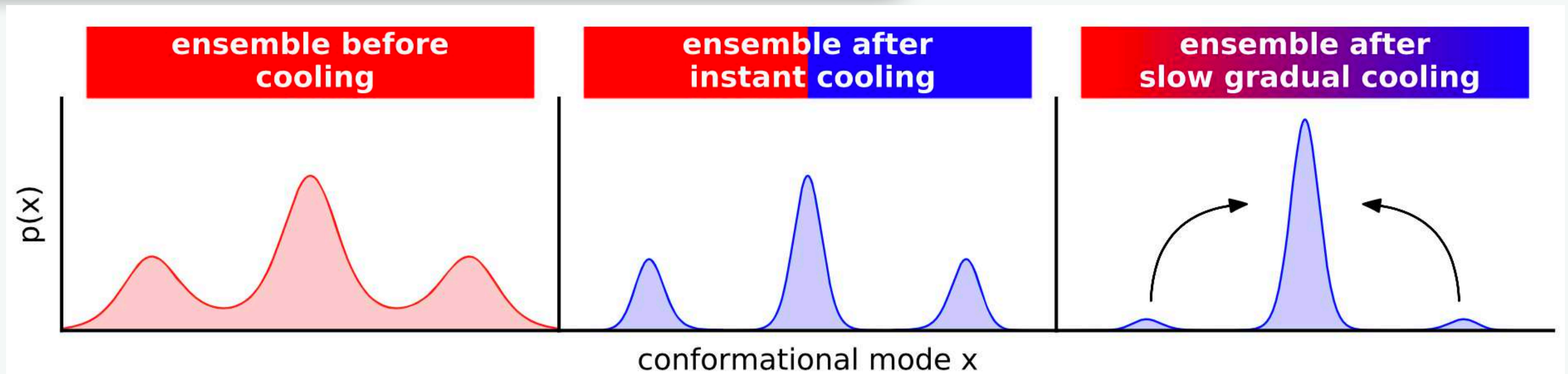
ARTICLE Check for updates

<https://doi.org/10.1038/s41467-022-29332-2> **OPEN**

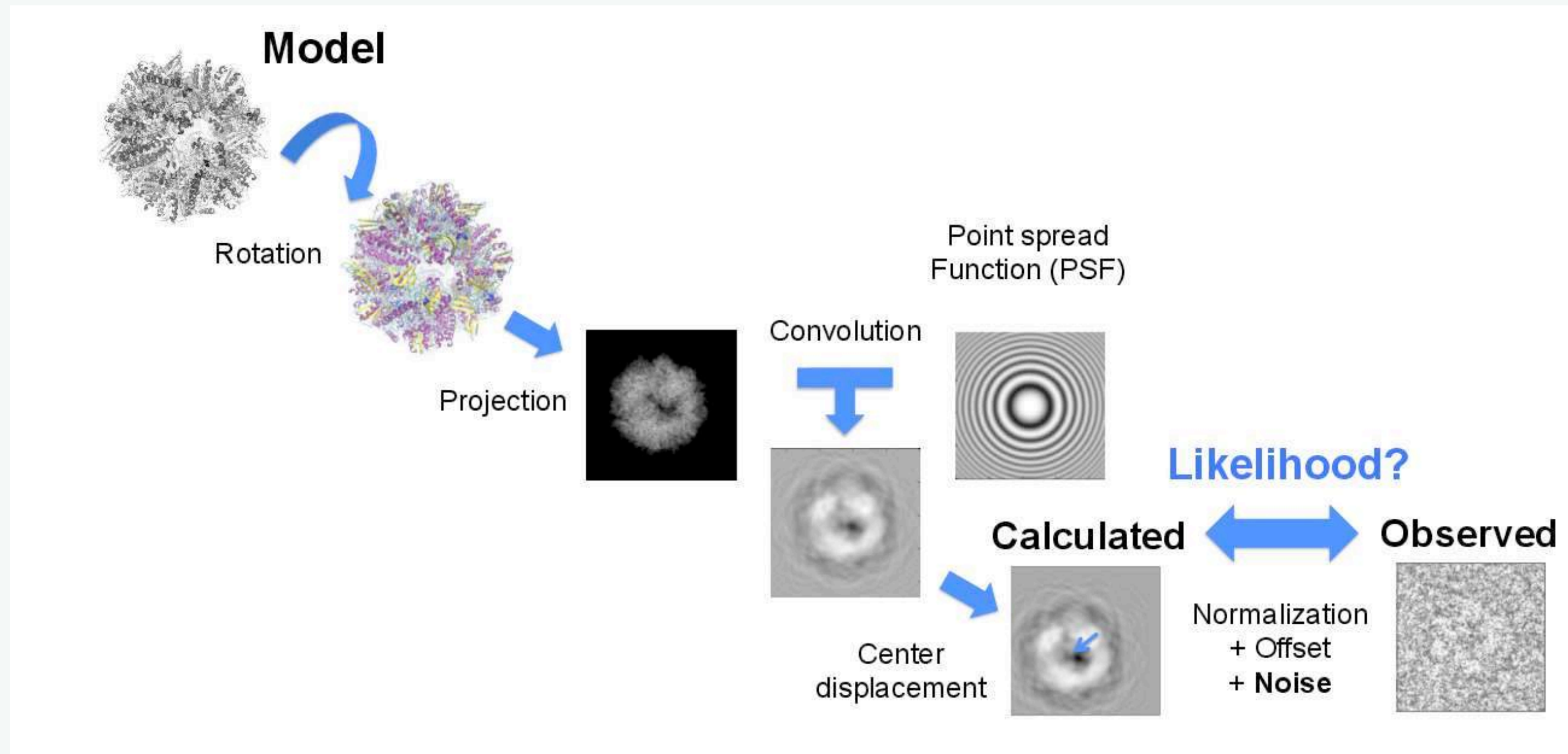
Effects of cryo-EM cooling on structural ensembles

Lars V. Bock ¹✉ & Helmut Grubmüller ¹

NATURE COMMUNICATIONS | (2022)13:1709 | <https://doi.org/10.1038/s41467-022-29332-2> | www.nature.com/naturecommunications

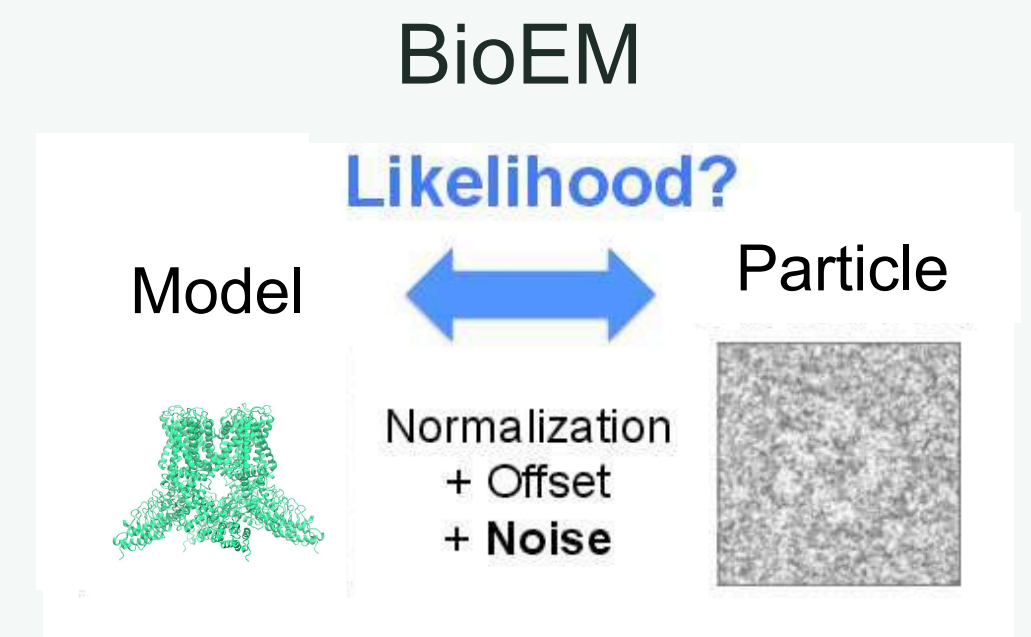
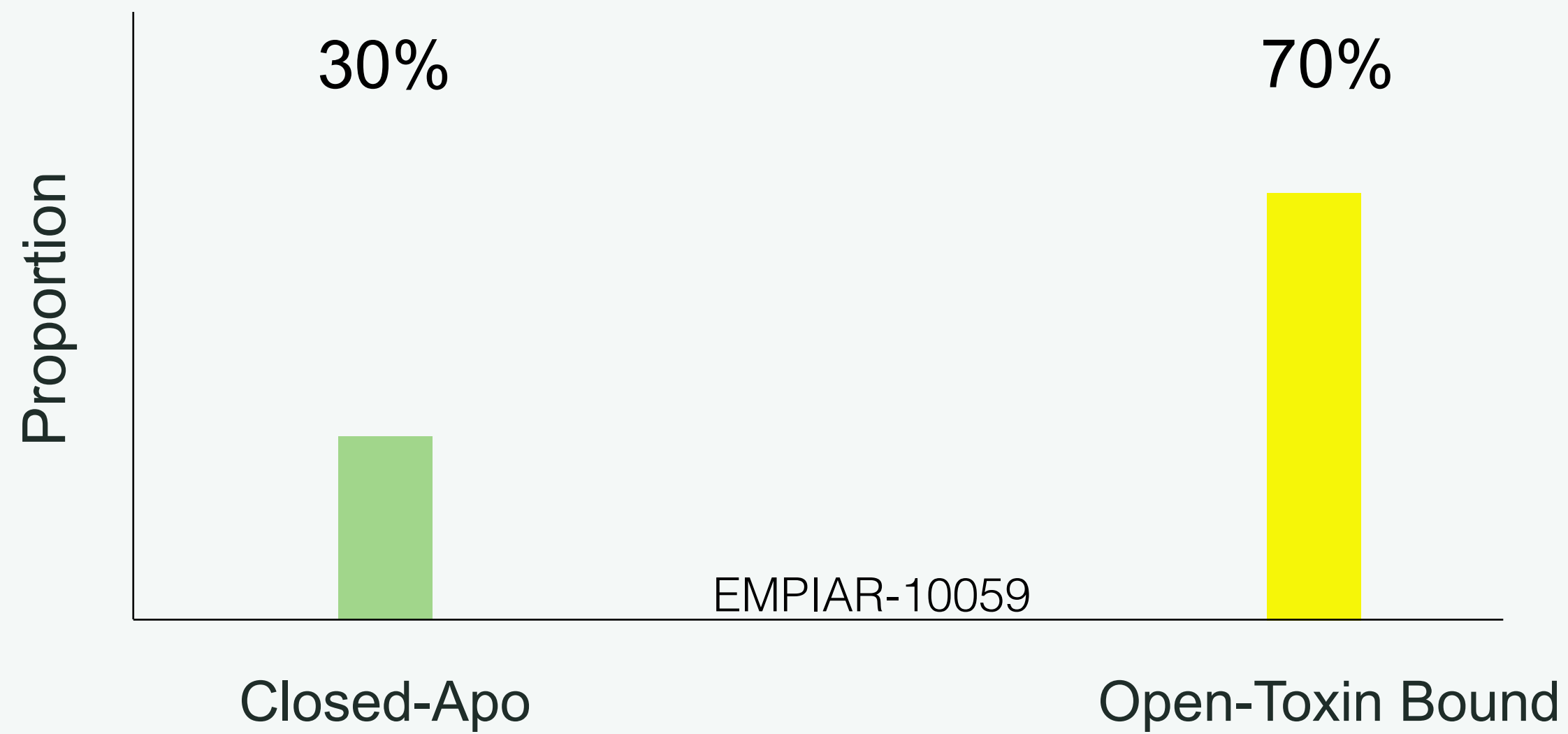
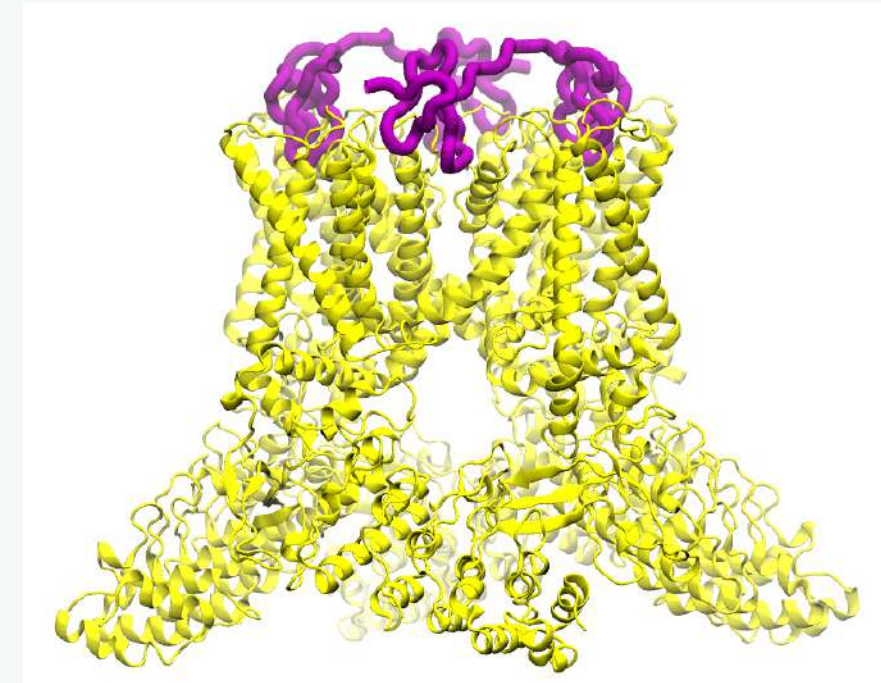
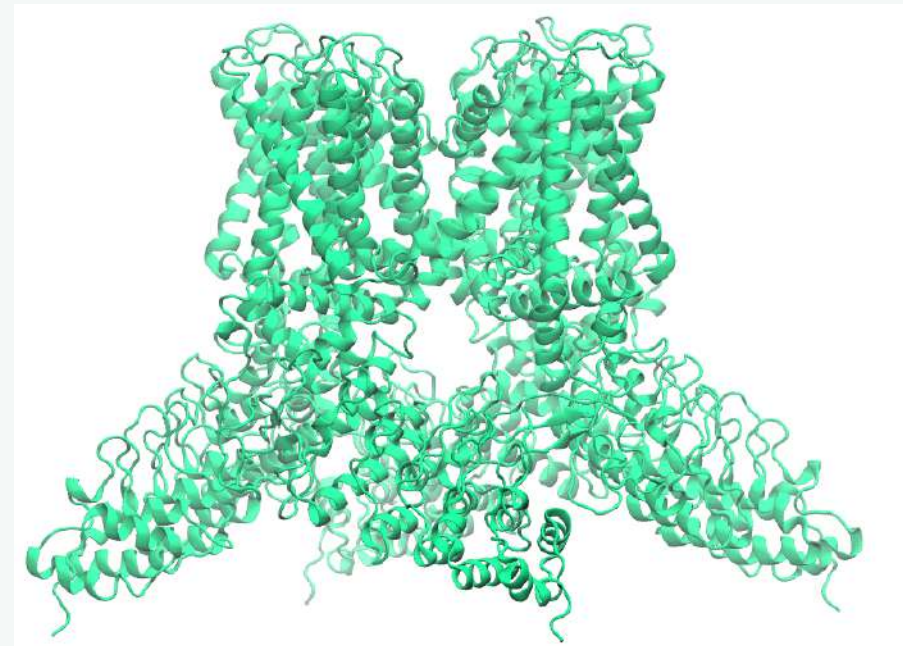


Bayesian Inference Methods For Comparing Single Particles to Atomic Models



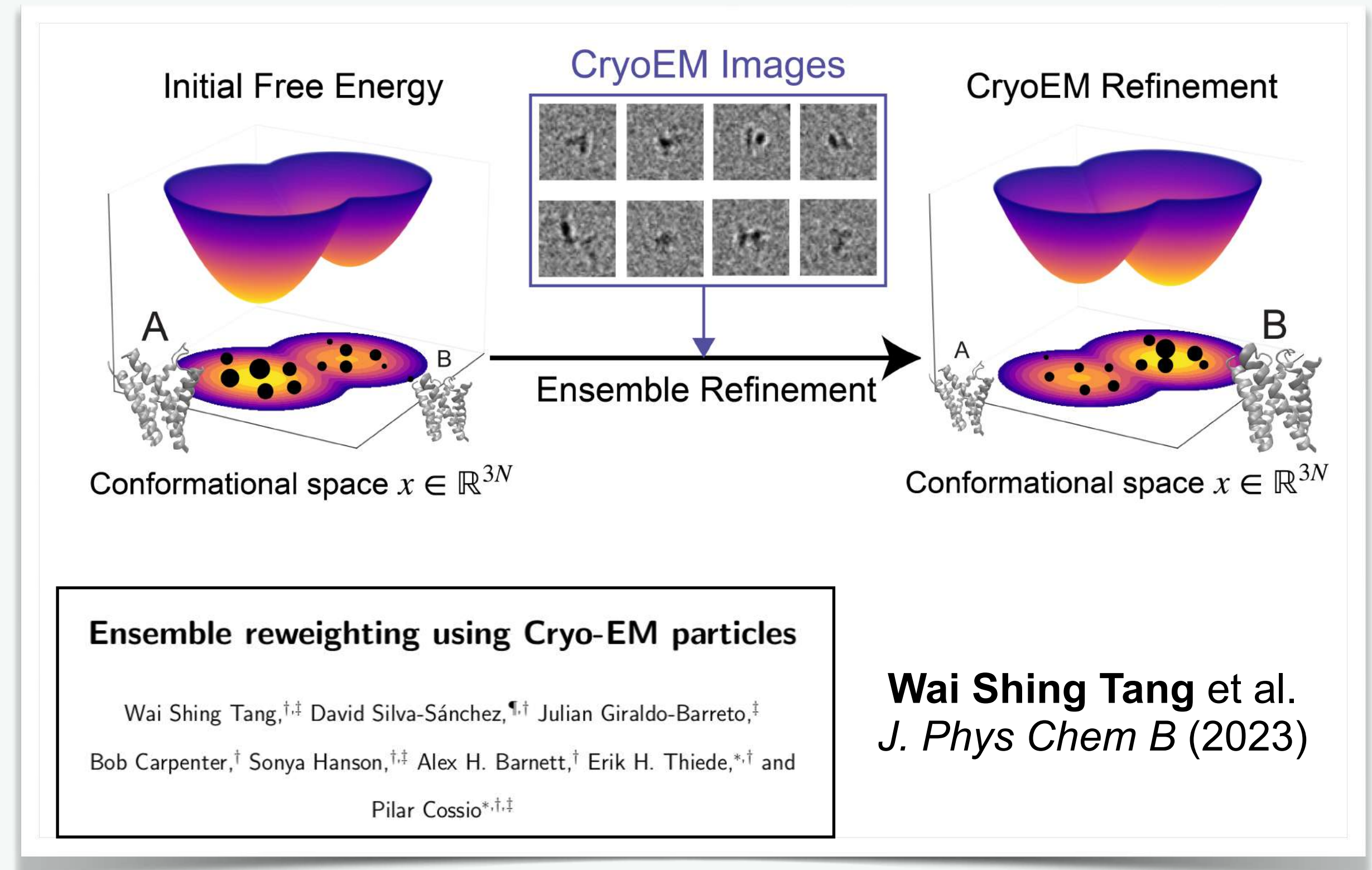
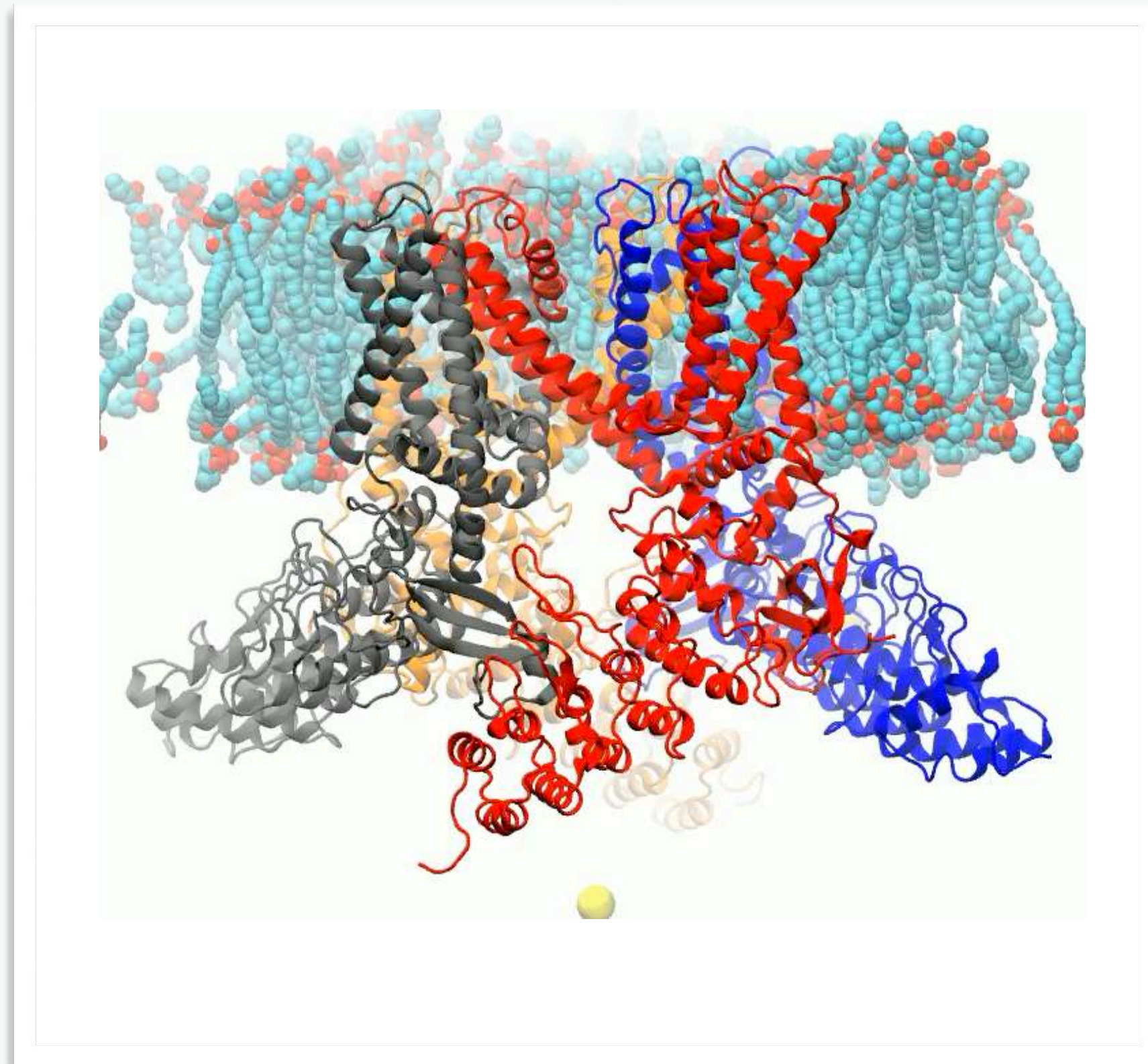
Pilar Cossio et al. Computer Physics Communications (2017)

Using Bayesian Inference to Classify Particles into Closed and Open States.



Cossio et al. Computer Physics Transactions (2017)

The future: Bridging the gap between cryo-EM data and long molecular dynamics simulations





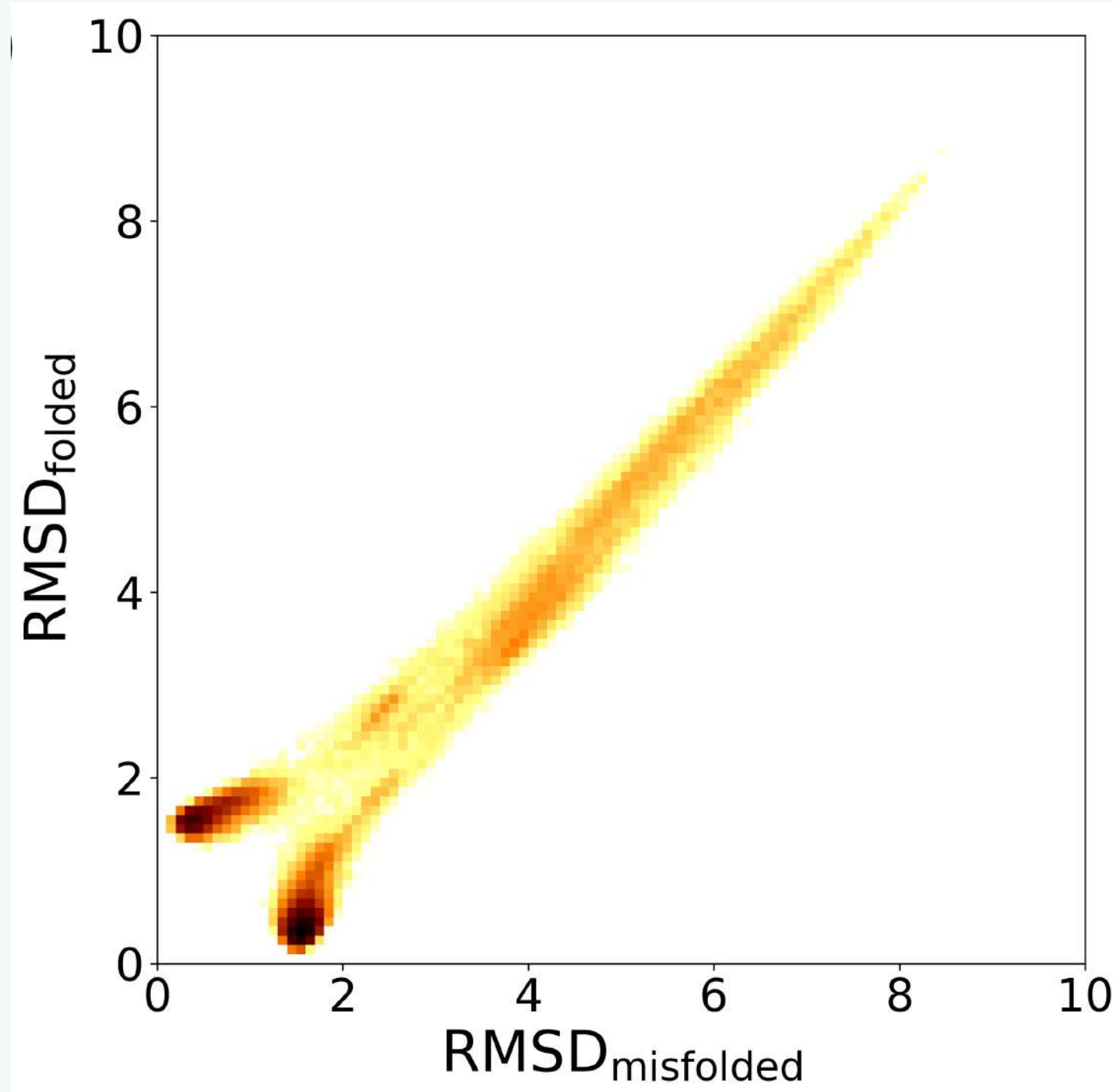
EH Thiede



WS Tang

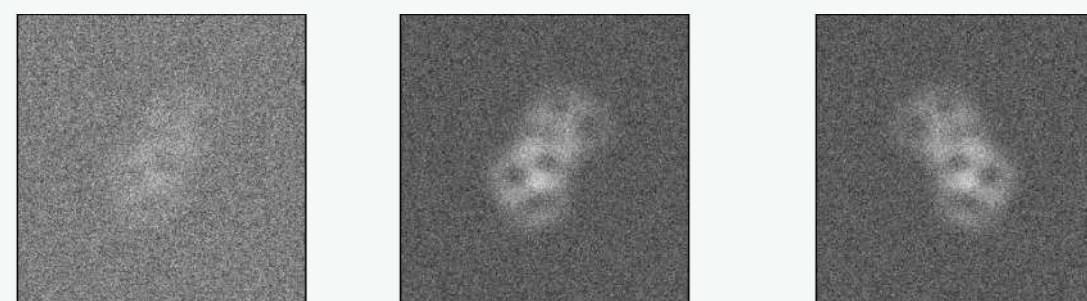
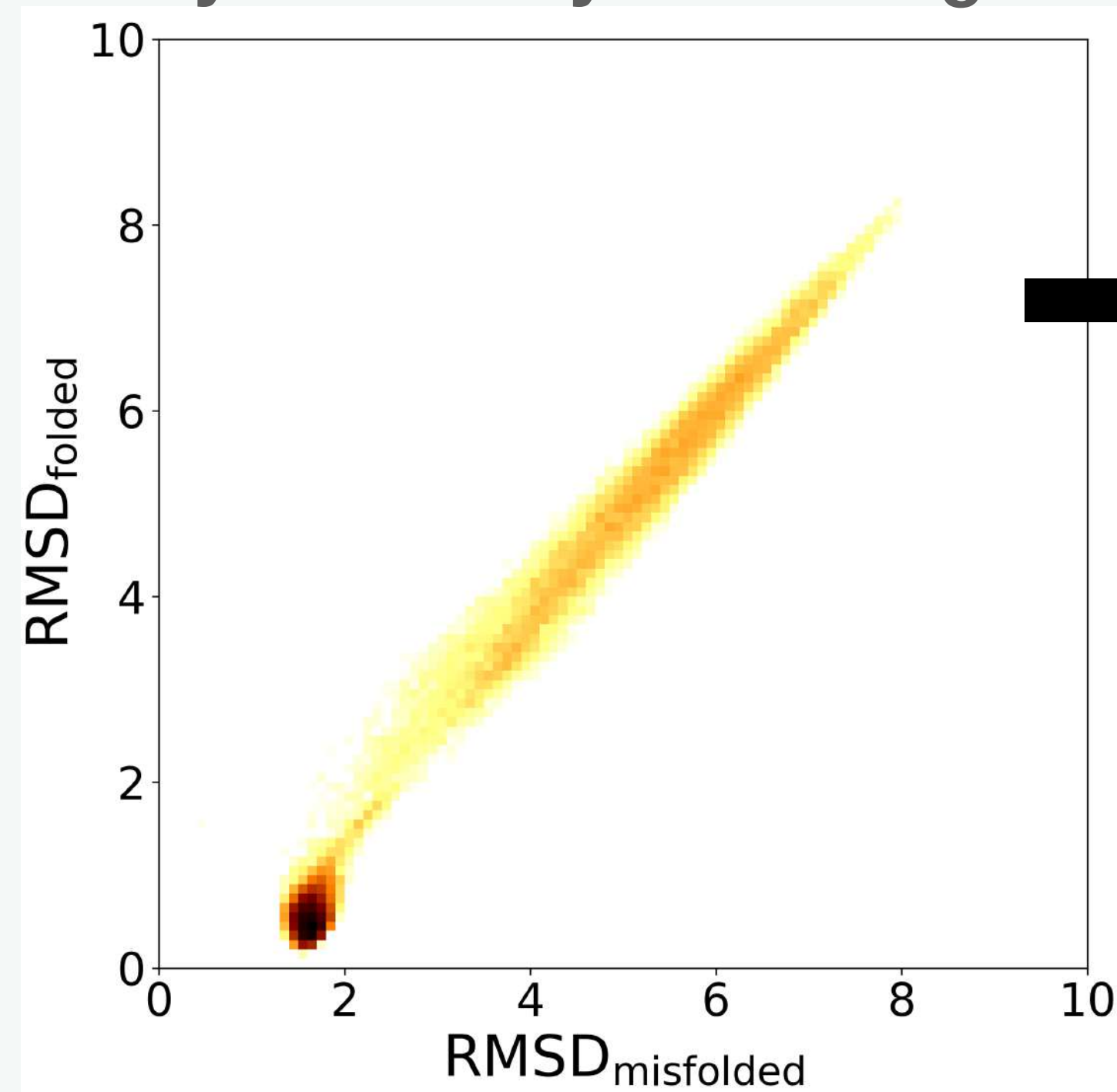
Cryo-EM ensemble reweighting

MD Structural Ensemble



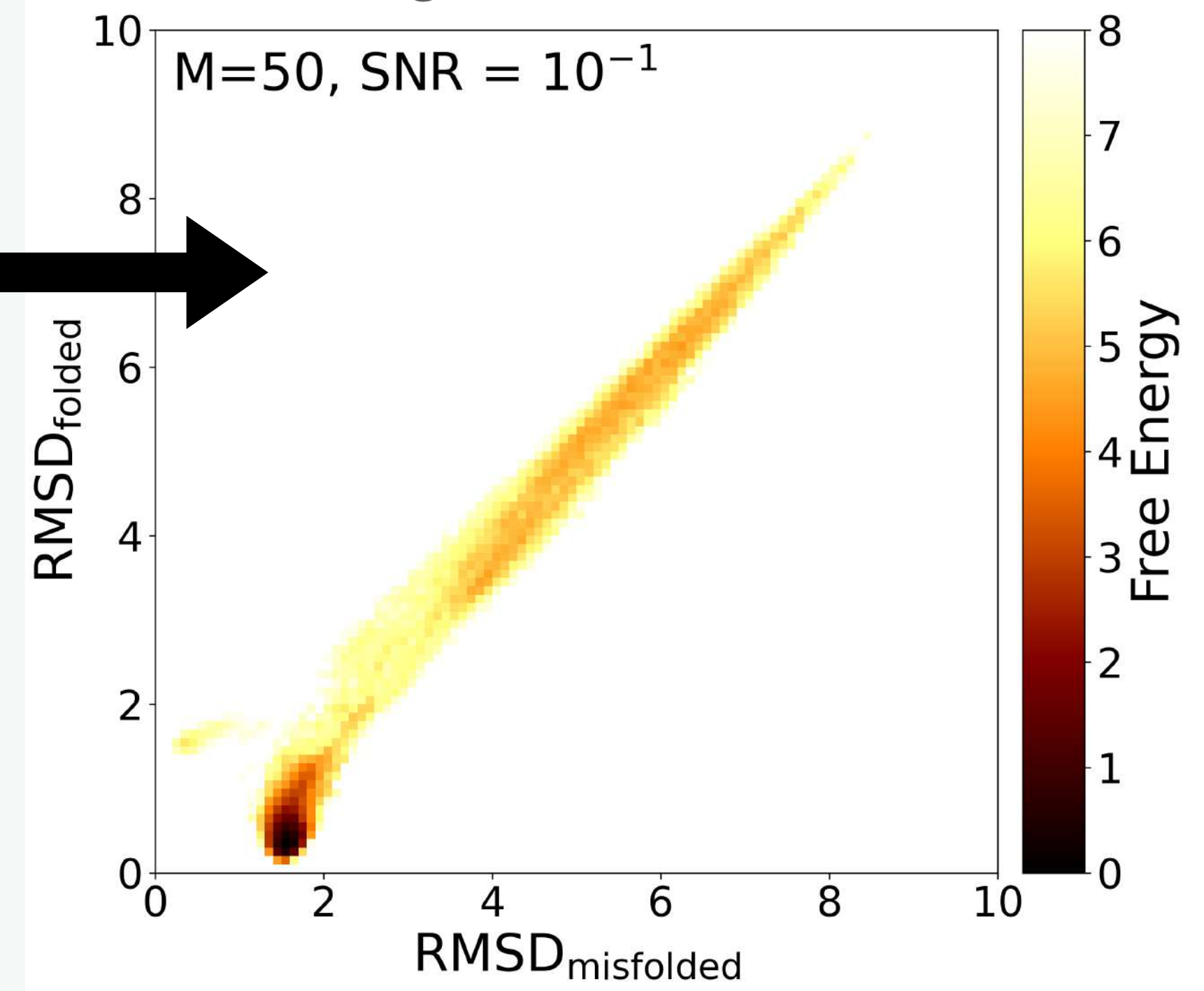
(our simulation)

Synthetic Cryo-EM Images



(DE Shaw data)

Reweighted Ensemble



MD Simulation reweighted by cryo-EM images



Assessing the performance of continuous heterogeneity methods.

Are these motions real?

462

IEEE TRANSACTIONS ON COMPUTATIONAL IMAGING, VOL. 8, 2022

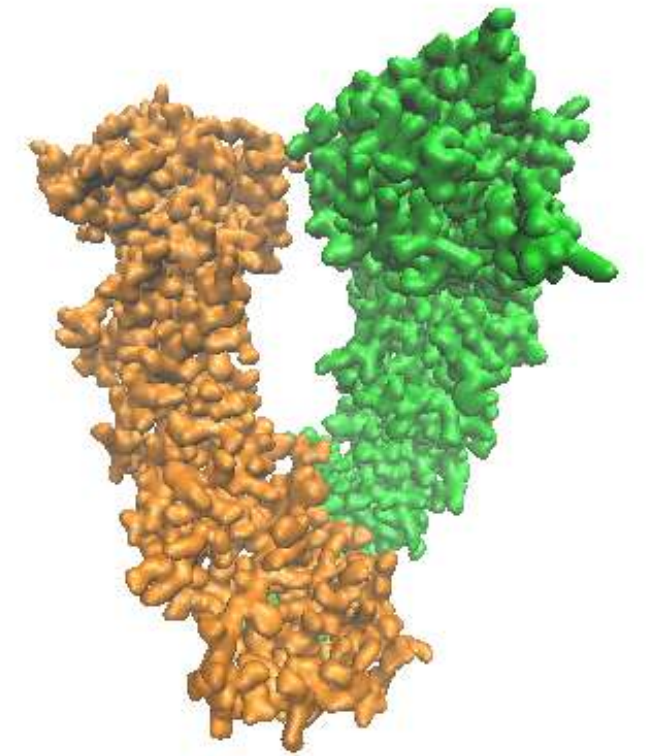
Recovery of Conformational Continuum From Single-Particle Cryo-EM Images: Optimization of ManifoldEM Informed by Ground Truth

Evan Seitz , Francisco Acosta-Reyes , Suvrajit Maji , Peter Schwander , *Member, IEEE*, and Joachim Frank 

Abstract—This work is based on the manifold-embedding approach to study biological molecules exhibiting continuous conformational changes. Previous work established a method—now termed ManifoldEM—capable of reconstructing 3D movies and accompanying free-energy landscapes from single-particle cryo-EM images of macromolecules exercising multiple conformational degrees of freedom. While ManifoldEM has proven its viability in several experimental studies, critical limitations and uncertainties have been found throughout its extended development and use. Guided by insights from studies with cryo-EM ground-truth data, simulated from atomic structures undergoing conformational changes, we have built a novel framework, ESPER, able to retrieve

I. INTRODUCTION

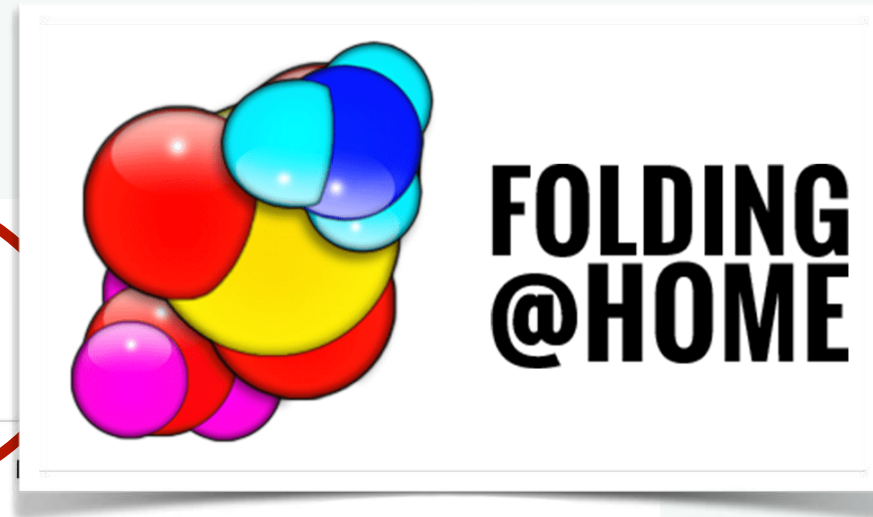
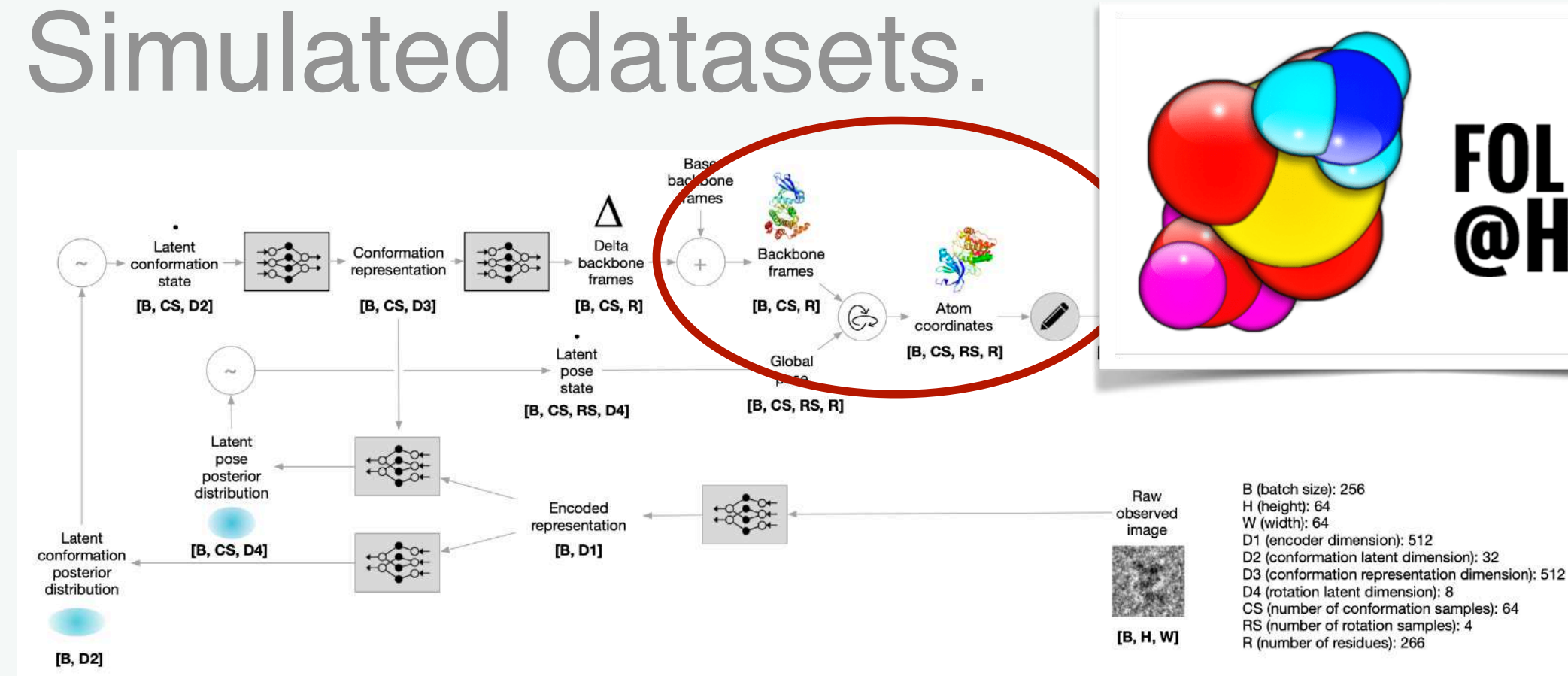
MOLECULAR machines—consisting of assemblies of proteins or nucleoproteins—take on a range of unique configurations or *conformational states* as they go through their functional cycles [1]. These states are typically characterized by different spatial constellations of relatively rigid domains, and can be organized in a *state space* according to the continuous motions of each domain along a unique coordinate. Specific sequences of the states in this space form pathways along which the molecular machine may transform. When the number of



Seitz et al *bioRxiv* (2019)
<https://doi.org/10.1101/864116>

An apoferritin for heterogeneity?

- Simulated datasets.



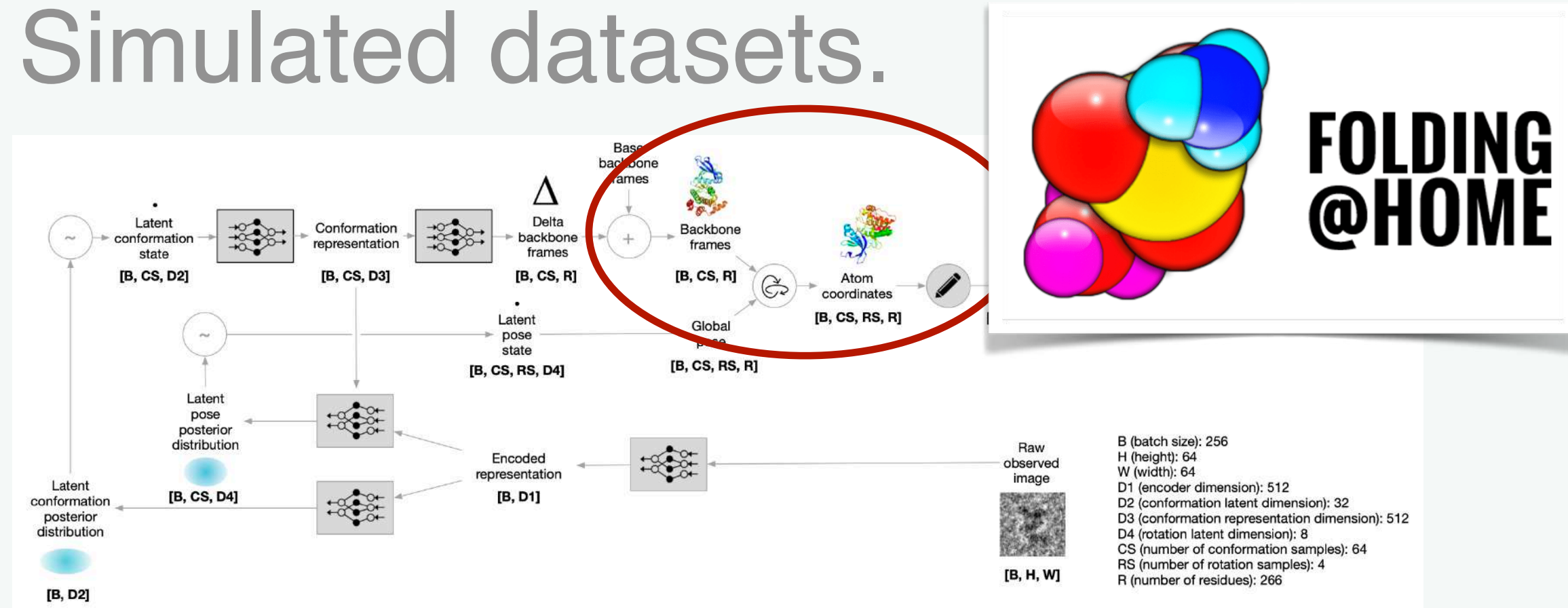
2021-06-25

Inferring a Continuous Distribution of Atom Coordinates from Cryo-EM Images using VAEs

Dan Rosenbaum^{*,1}, Marta Garnelo^{*,1}, Michal Zielinski^{*,1}, Charlie Beattie¹, Ellen Clancy¹, Andrea Huber¹, Pushmeet Kohli¹,
 Andrew W. Senior¹, John Jumper¹, Carl Doersch¹, S. M. Ali Eslami^{*,1}, Olaf Ronneberger^{*,1} and Jonas Adler^{*,1}
^{*}Equal contributions, ¹DeepMind

An apoferritin for heterogeneity?

- Simulated datasets.



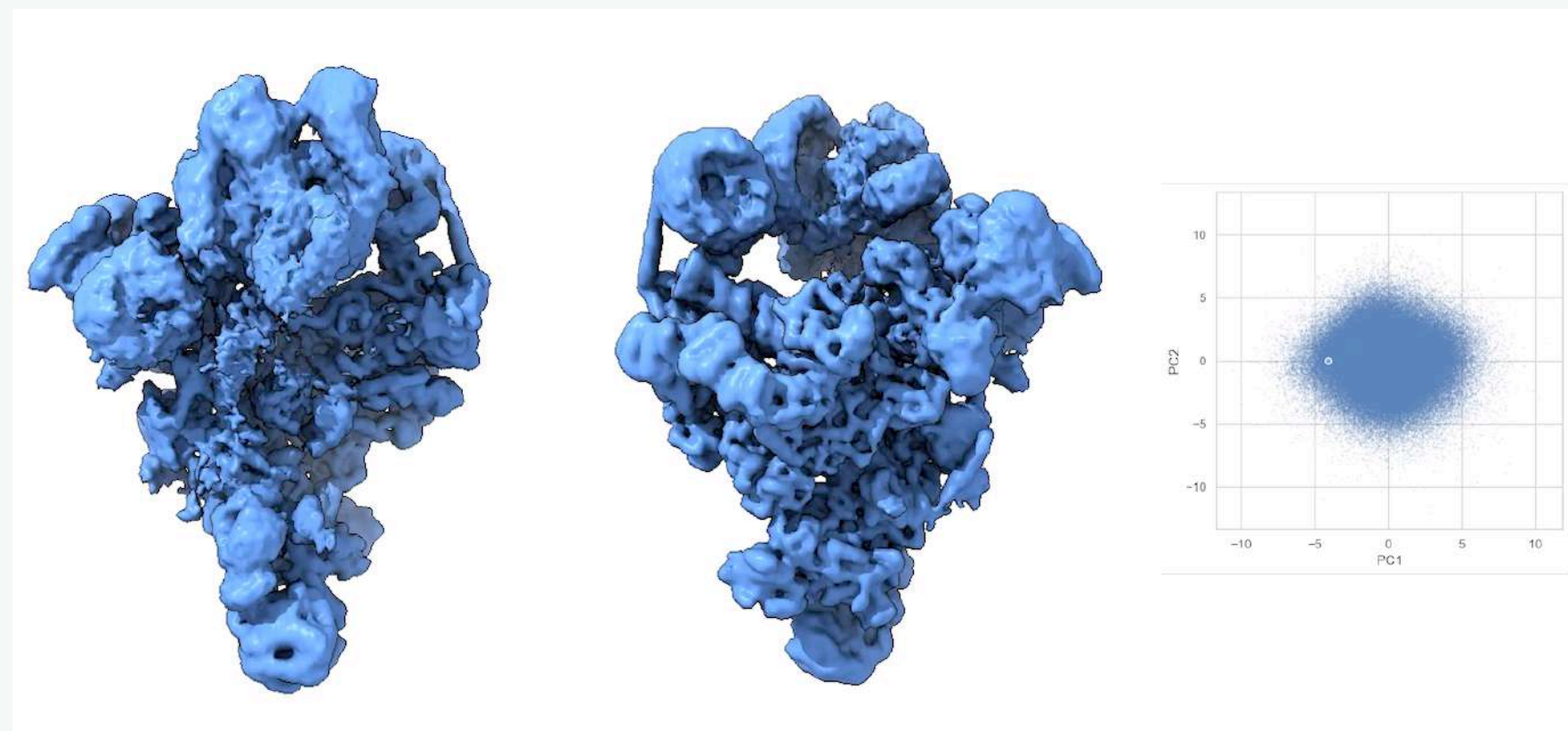
DeepMind 2021-06-25

Inferring a Continuous Distribution of Atom Coordinates from Cryo-EM Images using VAEs

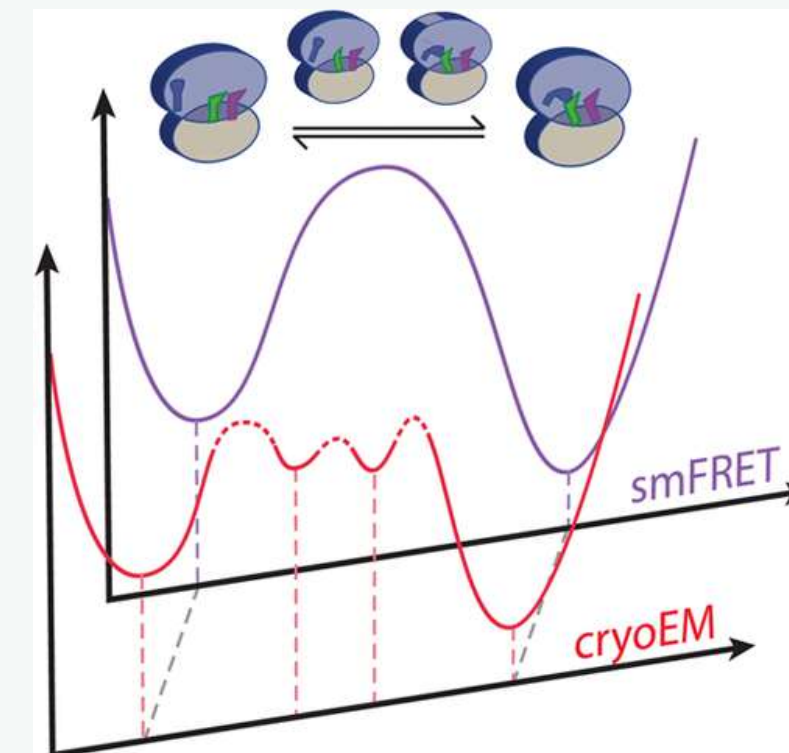
Dan Rosenbaum^{*,1}, Marta Garnelo^{*,1}, Michal Zielinski^{*,1}, Charlie Beattie¹, Ellen Clancy¹, Andrea Huber¹, Pushmeet Kohli¹, Andrew W. Senior¹, John Jumper¹, Carl Doersch¹, S. M. Ali Eslami^{*,1}, Olaf Ronneberger^{*,1} and Jonas Adler^{*,1}

*Equal contributions, ¹DeepMind

- Experimental datasets (w/ validation of the populations from another experimental method!)



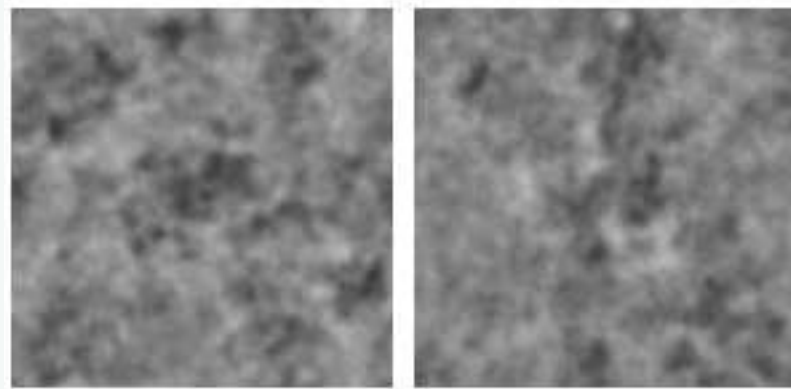
Zhong et al
Nature Methods (2021)



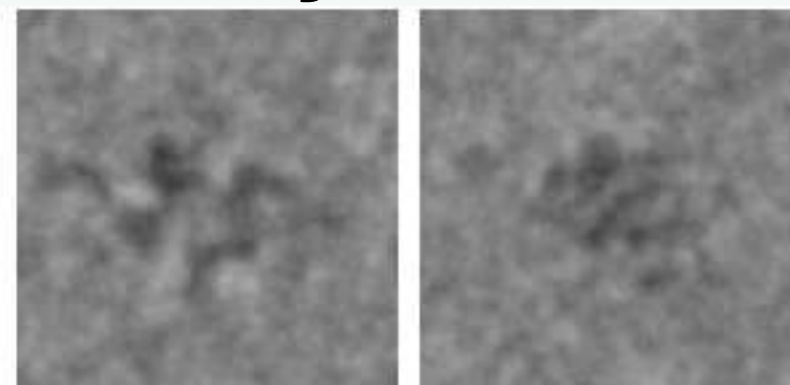
Kinz-Thompson et al
J Phys Chem B (2015)

Validating methods for heterogeneity in cryo-EM is still needed...

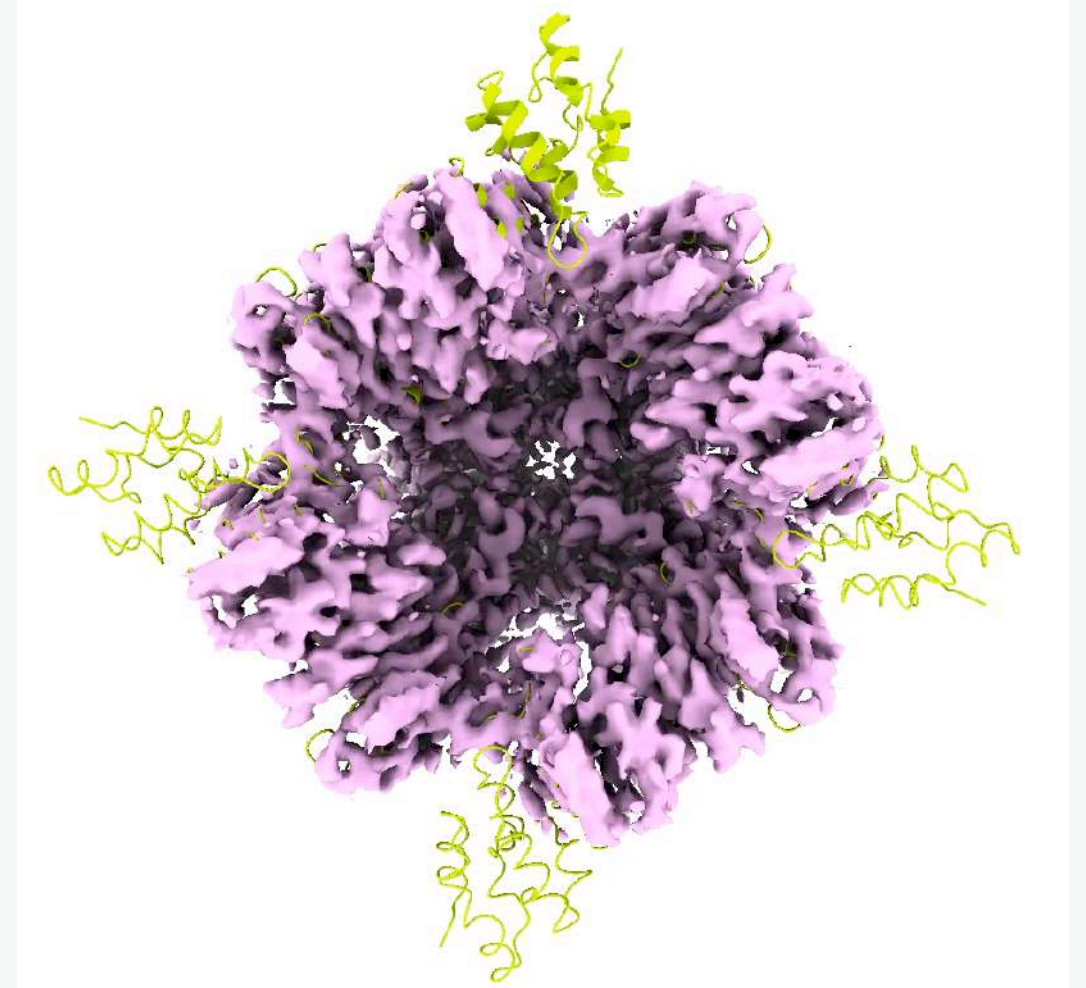
real



synthetic



There are a variety of these methods ranging from linear spectral methods to deep neural nets to reweighting of MD ensembles, and the each have different **strengths and weakness** that we are trying to tease out.



... especially on real experimental datasets.

Deep generative modeling for volume reconstruction in cryo-electron microscopy

Claire Donnat^{a,1}, Axel Levy^{b,c}, Frédéric Poitevin^c, Ellen D. Zhong^d, Nina Miolane^{e,1}


- (i) **Lack of benchmark datasets:** Current methods are developed and tested on a wide range of synthetic and experimental datasets that differ in the nature of the biomolecule being imaged, the dataset size, image size and associated resolution — with very little overlap across methods - see Table summarizing existing experiments in the supplementary material. There is unfortunately no MNIST (Deng, 2012) or Imagenet (Deng et al., 2009) for cryo-EM. Most methods resort to evaluating their performance on synthetic data, yet no cryo-EM simulator acts as a standard to generate simulated images in a unified way. Synthetic datasets vary in the realism of the image formation model used for simulation, e.g. in the noise model, the signal-to-noise ratio or the distribution of nuisance variables (e.g. poses). Subsequent ex-

2023: A blind heterogeneity challenge for cryo-EM hosted by the Flatiron Institute.

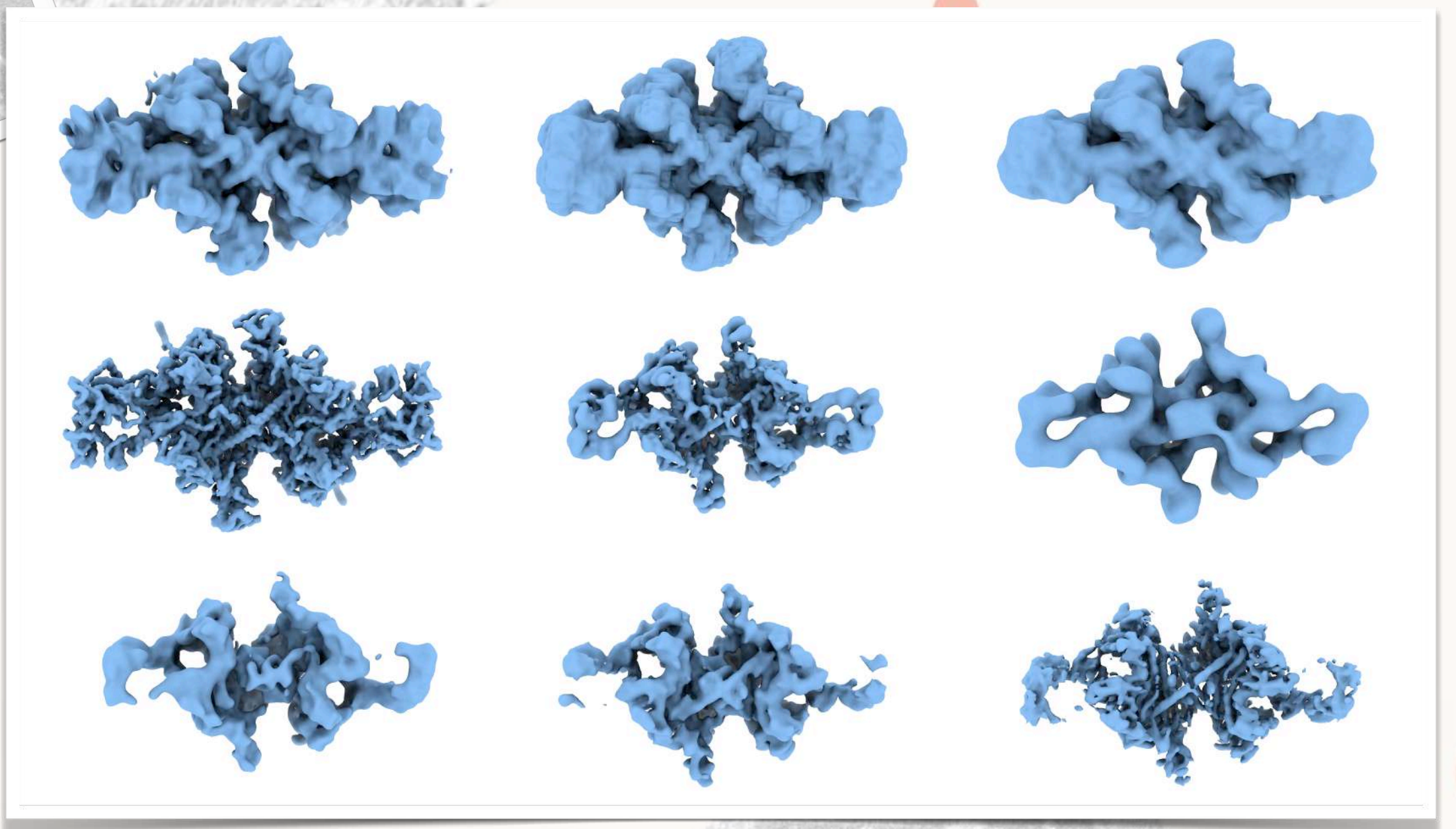
Average

Participate by analyzing the two heterogeneous datasets provided and uploading your results between September 1 and October 31, 2023. More info at:

(Note that this is a challenge to explore how to compare methods in the first place and no winner will be declared. We welcome feedback/questions/comments!)

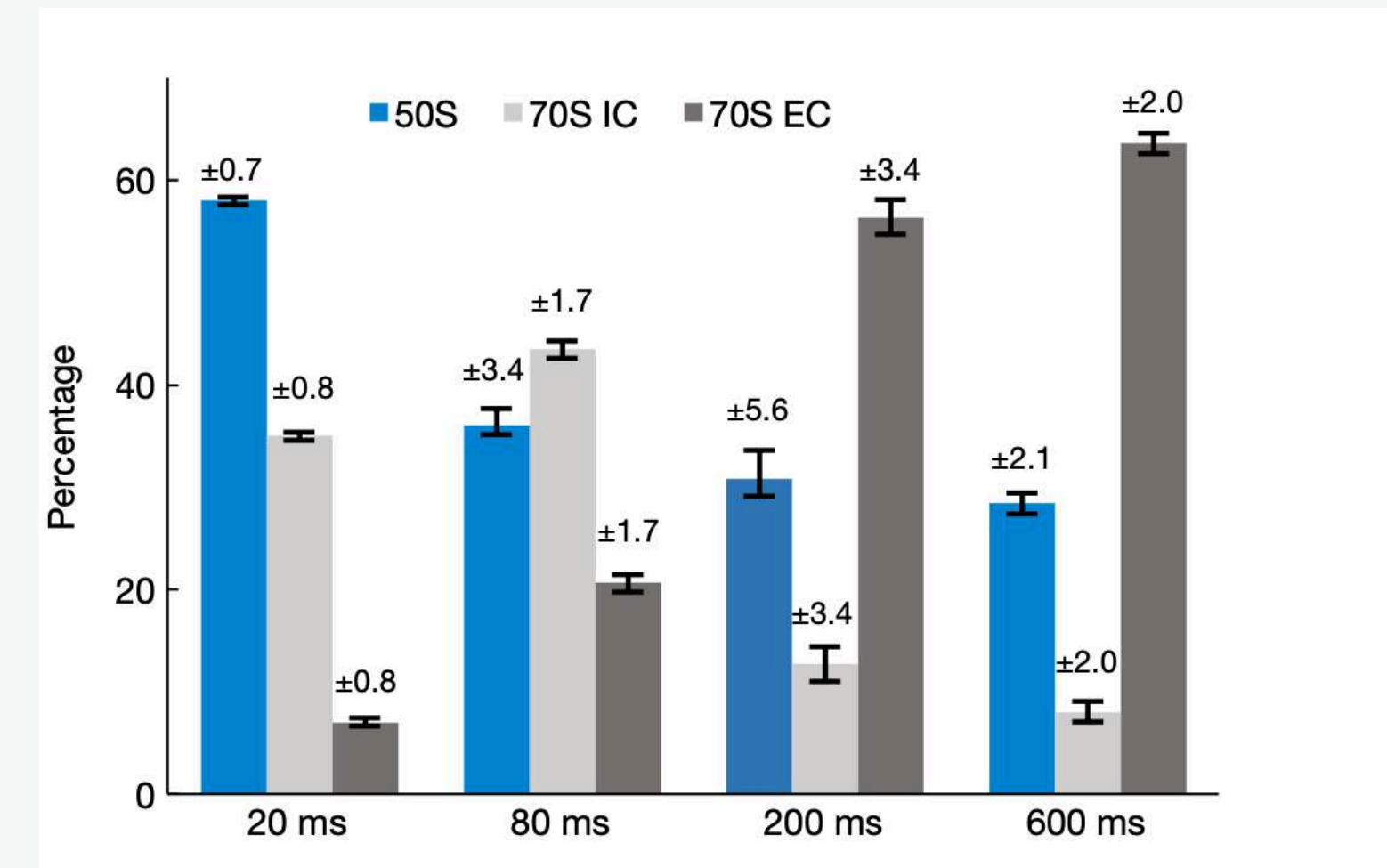


Two Datasets:  Average
1- real from NYSBC
2- simulated via MD, so we have ground truth states and their distribution



One potentially valuable aspect of continuous heterogeneity methods is getting populations, which can be translated into Free Energy landscapes.

- Quantifying errors in populations of states - also relevant to discrete heterogeneity.



Kaledhonkar et al, *Nature* (2019).

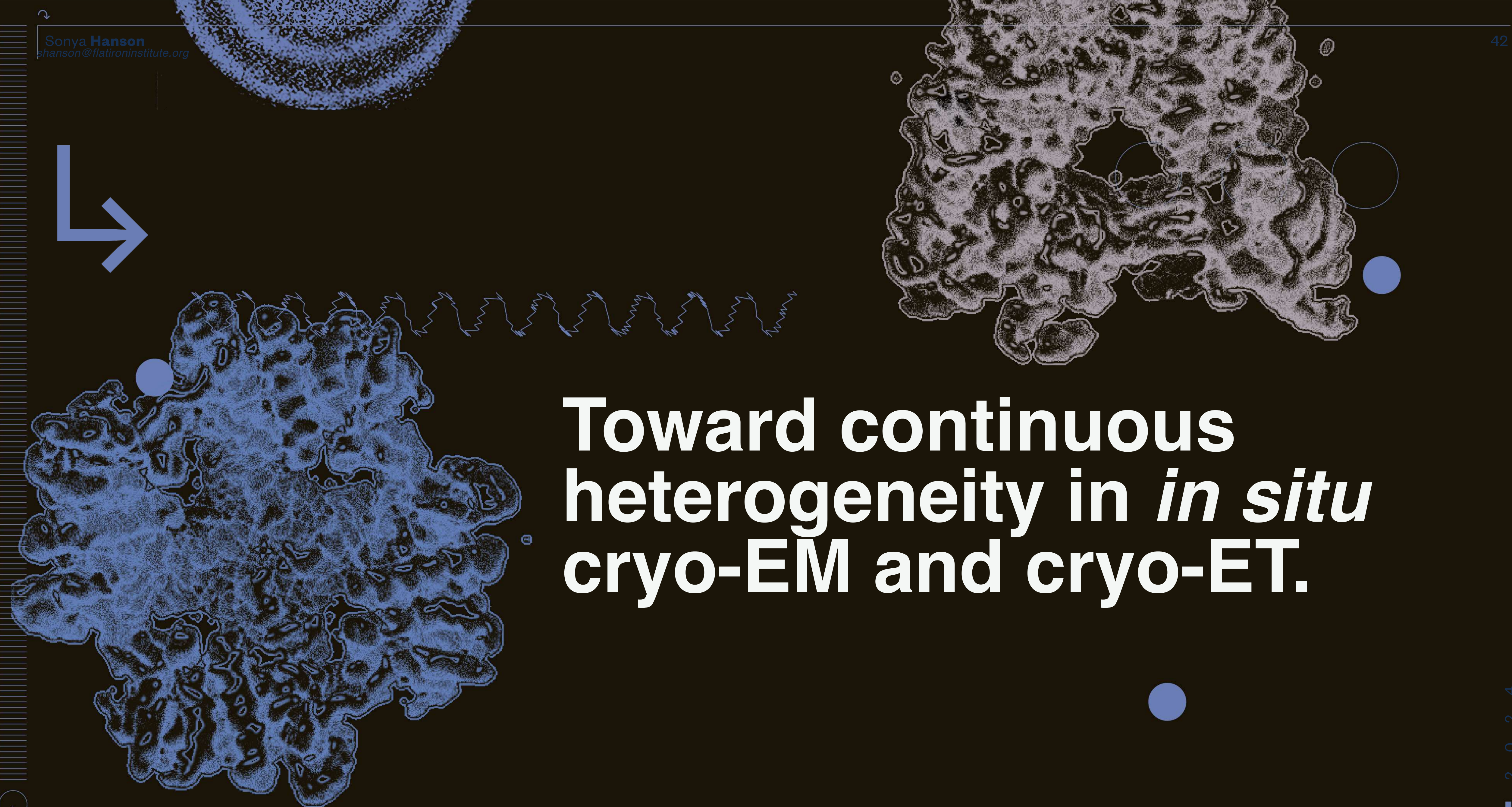
DISCUSSION TOPICS

How should we benchmark methods in continuous heterogeneity for cryo-EM?

How much should we prioritize getting populations?

How much do we think the vitrification process affects our ensembles?

et cetera



Toward continuous heterogeneity in *in situ* cryo-EM and cryo-ET.