Cryo-EM, part. IV

From low-res blob... to atomic glory(-ish)

(and the many pittfalls to avoid)



Single particle analysis :

- Workflow overview
- How to obtain an initial, low res volume
- Refinement, validation, interpretation of cryo-EM 3D maps

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Single particle analysis : General principles

TEM images are 2D projections of randomly oriented 3D objects



Single particle analysis : workflow



de la Rosa-Trevin et al., 2016, DOI: http://dx.doi.org/10.1016/j.jsb.2016.04.010

First 3D reconstruction : how do we get an initial, low res 3D volume?



Tomography-like approaches :

"Tomography allows to reconstruct essentially everything which generates some sort of contrast."

(Andy Hoenger, EMBO course on 3D cryo-EM, Heidelberg, August 2004)



On the plus side...

No prior structural knowledge required

On the minus side...

Technically challenging (the smaller the object, the more challenging) RCT and ORT almost always realized on negatively stained preps Missing wedge/missing cone => deformation can happen

First 3D reconstruction : how do we get an initial, low res 3D volume?



Common lines approaches (angular reconstitution)

Central section theorem : The 2D Fourier Transform of the projection of a 3D object is a central section (a section passing through the origin) of the 3D Fourier transform of this object.



First 3D reconstruction : how do we get an initial, low res 3D volume?





Probabilistic approaches Stochastic gradient descent (SGD) : Relion, cryoSPARC



Punjani et al., Nature Methods2017

Pros : No prior structural knowledge required Does not require user intervention/(in)experience Very fast method CryoSPARC can sort out 3D classes at this step

Cons : (Relion)

even distribution of viewing directions data good enough to yield detailed 2D classes

Single particle analysis : workflow



Orientation determination by projection matching

Particles are compared and aligned to all the 2D projections of an existing 3D model Orientation is given by the reprojection giving the best correlation coefficient with the considered particle

Data Overfitting / Model Bias (aka « Einstein from noise »)

« Ideal » case

White noise

Steve Ludtke

Wrong model

Bad model in => bad model out

Human RNA Pol II PIC

Yeast RNA Pol II PIC

He et al. & Nogales, Nature 2013

Murakami et al. & Kornberg, Science 2013

Orientation determination: PM + probabilistic approaches (maximum likelihood)

Particles orientation are given by projection matching, but a weighting is applied to each particle before 3D reconstruction. "Good" particles have more weight than "bad" ones. This can at least partly prevent overfitting.

Leschziner and Nogales, 2007. DOI: 10.1146/annurev.biophys.36.040306.132742

3D Refinement

Iterative ML projection matching from an initial, low resolution model Goal : align and determine orientation of particles with the highest possible precision degree At the beginning, the dataset is randomly divided into two groups, refined separately from each other The 3D reconstruction resulting from a cycle of MLPM will be used as reference for the next one

At the end of 3D refinement

Calculate a full (combined half datasets) map Estimate map resolution Low-pass filter the full map according to resolution

Unfiltered half map1

Full filtered map

Resolution Estimation: Fourier Shell Correlation

The aligned & oriented particles dataset is randomly divided in two, and two 3D reconstructions are calculated

Both 3D independent "half maps" are compared in Fourier space

The resolution of the whole 3D reconstruction corresponds to the spatial frequency where correlation coefficient between half maps drops below a value X

X was 0.5 for a long time, and has been lowered to 0.143, according to Rosenthal et Henderson, *j mol biol* 2003

C_{Ref}: coef. between a perfect map without noise and an experimental map reconstructed from the full dataset

Refined 3D Map : what now?

- Local 3D Classification / Refinement
- Map sharpening (global/local) : amplitudes correction so that low-res & high-res terms have correct relative scaling

Filtering vs. sharpening

Unfiltered half map1

Full filtered map

Full filtered+sharpened map

Map oversharpening effects

Murshudov, 2016

Refined 3D Map : what now?

- Local 3D Classification / Refinement
- Map sharpening (global/local) : amplitudes correction so that low-res & high-res terms have correct relative scaling

True for maps @10 Å resolution and better !

• Map Interpretation : fitting techniques depends on map resolution

20 Å	15 Å		10 Å	5 Å	2 Å
	Rigid Body Docking		Flexible Fitting	<i>de n</i> moc	ovo atomic del building
Global shape Occupancy		Molecule orientation	a helices	β strands	lateral chains

Hétérogénéité structurale / états conformationnels

Gamma secretase : disorder in the catalytic site of presenilin

Bai et al., 2015 DOI: http://dx.doi.org/10.7554/eLife.11182

Single particle analysis : breaking the 100kDa

Koshouey et al., 2017

Hemoglobin (64 kDa)

Contraste très faible en cryo-TEM : il était difficile de visualiser des objets inférieurs à 5 nm de diamètre (~150 kDa). Nouvelles technologies (caméra à détecteur direct d'électrons, phase plate) permettent de visualiser des objets de plus en plus petits (< 100 kDa)

Microscopies Corrélatives

Figure 10 Cryo-CLEM imaging of retroviral endocytosis and fusion. (**a**-**c**) Double-labeled HIV-1 particles pseudotyped with avian sarcoma and leukosis virus (ASLV) Env glycoprotein. (**a**) Cryo-CLEM of ASLV Env pseudotyped HIV-1 particles bound to CV-1/TVA950 cells. Central region is the overlay of the cryo-EM montage onto the cryo-fLM image. Red square indicates the tomography data in **b** and **c**. (**b**) Tomographic slice, with segmentation, of ASLV Env pseudotyped HIV-1 particles undergoing endocytosis. The viral membrane (light blue) and mature core (yellow) are rendered. Clathrin cages (purple) surround several viral particles. (**c**) Enlargement of one clathrin cage (purple) surrounding a viral particle (light blue). Scale bars, (**a**) 50 µm, (**b**) 250 nm, (**c**) 100 nm.

Tomographie électronique

Medalia et al., 2002

Subtomogram averaging

Parts of tomograms are extracted, aligned and averaged using SPA approaches

Fig. 1. Structure of the immature HIV-1 CA-SP1 lattice at 3.9 Å. (**A**) Computational slices through tomograms of immature HIV-1 (D25A mutant), untreated ΔMACANCSP2 VLPs, and BVM-treated ΔMACANCSP2 VLPs. The arrowhead marks the membrane bilayer in the left panel. Scale bar, 50 nm. (**B**) Electron densities of CA-SP1 from the samples shown in (A), viewed perpendicular to the lattice, generated by subtomogram averaging. One CA-SP1 monomer is highlighted in color, with the CA-NTD in cyan and the CA-CTD and SP1 in orange. The resolutions of the determined structures are noted. (**C**) The refined atomic model, viewed from outside of the virus (top) and rotated by 90°, shown in the same view as in (B) (bottom). The sixfold symmetry axis is indicated with a hexagon.

Schur et al., Science 2016 DOI: 10.1126/science.aaf9620

Tomography-like approaches : random conical tilt, orthogonal tilt reconstruction

2 images of the same particles field, one tilted compaired to the other

From A. Leschziner, https://software.rc.fas.harvard.edu/leschziner public/index.php/Main Page

Tomography-like approaches

Random conical tilt

With only two exposures a conical tilt series can be generated

Angular distribution represented on a spherical angular map

Probabilistic approaches Stochastic gradient descent (SGD) : Relion, cryoSPARC

Optimization objective function: full likelihood using all images

Space of all 3D structures