

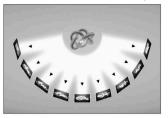


- Random Conical tilt 3D reconstruction
- Euler angles
- · Principle of conical tilt series

· Central section theorem

- · Missing cone artefact
- · Multivariate statistical analysis
- Early 3D studies and negative staining problems
- · Perspectives and new trends

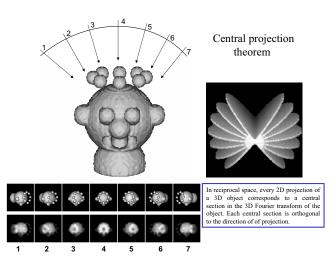
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Tilt series



Back-projection & 3D reconstruction



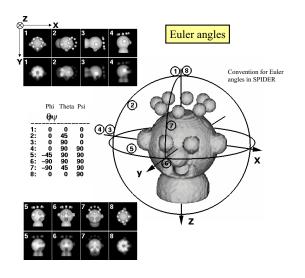
Constraints of cryoEM on biological objects:

Work with low electron dose (~10e-/Ų) => the less exposures, the better.

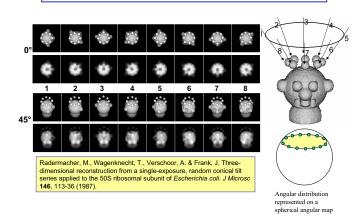
Images have a low signal-to-noise ratio
Compromise defocus level with resolution (CTF)
Computing 2D or 3D numeric averages

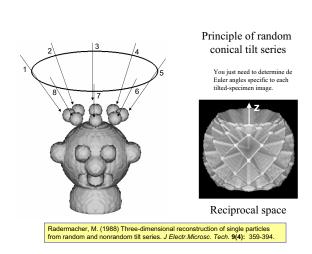
→ (only one conformation assumed in the sample)

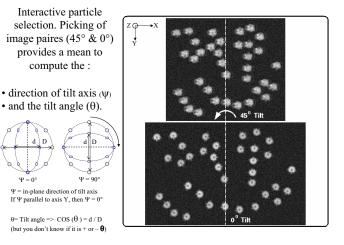
Use internal symmetries of the objects: helicoïdal symmetry, icosaedral symmetry, 2D crystals, or no symmetry at all...



With only two exposures a conical tilt series can be generated

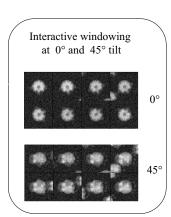






2D projections are identical, except for an in plane rotation corresponding to Euler angle ϕ .

> 2D projections are not identical due to the tilt. Moreover, neighboring particles start to overlap



Rotation of each 0° projection by its -φ angle

Interactive particle

provides a mean to

compute the:

• and the tilt angle (θ) .

in-plane direction of tilt axis If Ψ parallel to axis Y, then $\Psi = 0^{\circ}$

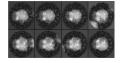
 θ = Tilt angle => COS (θ) = d / D

A circular mask hides (up to a certain point) the neighboring particles.

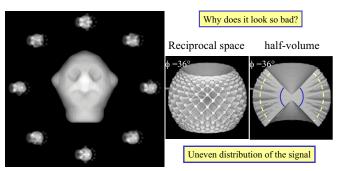
2D alignment of untilted-specimen images and computation of angle ϕ



Centering and masking of tiltedspecimen images

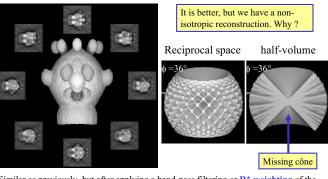


Simple back-projection



Once the 3 Euler angles are determined, the 3D reconstruction can be performed from the tiltedspecimen projections. The simple back-projection is nothing more than adding in reciprocal space the FT of the 2D projections in their relative orientations (waffle-like distribution of central sections), followed by Fourier transform of this 3D distribution to come back in real space.

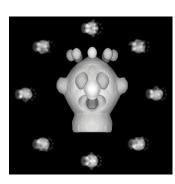
Weigthed back-projection



Similar as previously, but after applying a band-pass filtering or R* weighting of the signal (lowering contribution in low spatial frequencies).

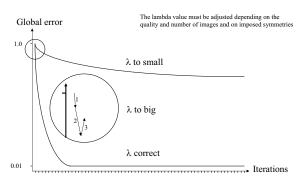
Simultaneous Iterative reconstruction techniques (SIRT)

Real space & iterative:



In real space with iterative methods, a starting volume is computed by simple back-projection. Then, the volume is reprojected in its original directions and 2D projection maps are compared with the experimental EM images. The difference maps [(EM) minus (2D projection of the volume)] are computed and back-projected to correct the 3D reconstruction volume. To avoid "overcorrecting" the structure, the 2D difference maps are multiplied by an attenuation factor λ , (with $\lambda \sim 0.5 \cdot E-04$ to 0.1.E-06). This process is iterated and at each step the "global error" between EM images and the computed volume is measured to check improvement.

Correct λ value?



Comparing 3D reconstruction techniques

Original object

Weigthed back-projection



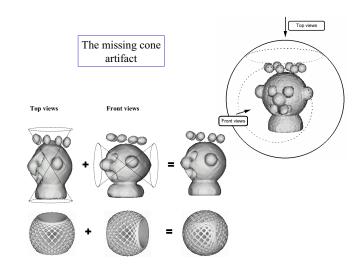


Simple back-projection

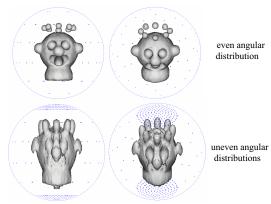




SIRT

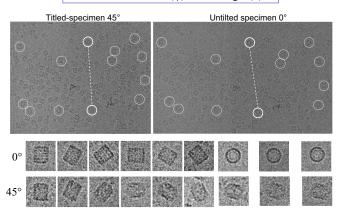


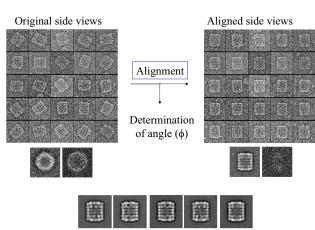
In rare occasions, a single overabundant preferential orientation can distort your structure when using SIRT



Boisset N., Penczek P., Taveau J.C., You V., de Haas F., Lamy J. (1998) Overabundant single-particle electron microscope views induce a three-dimentional reconstruction artifact. *Ultramicroscopy*, 74: 201-207.

Interactive particle picking with determination of tilt axis direction (ψ) and tilt angle (θ)





× 0 0 0 0 0 * × e

A simplified and therefore mathematically incorrect description of Correspondence analysis (CORAN)

You have normalized, aligned a set of noisy images and you want to sort them automatically. (For correspondence analysis no negative density is tolerated, while for principal component

MSA and clustering \rightarrow 5 views

"Intelligenti pauca" = intelligent people understand each other with a few words!

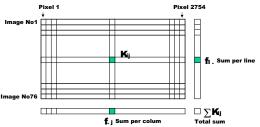
1- Create a mask following the shape of the total average







- 2- For each image, extract all densities from the pixels falling within the mask and re-dispose then into a line
- 3- Place theses lines of densities into a table



4- An other way to consider the data is to say that these densities are coordinates in a multidimensional space.
5- Hence in this example, each image having 2754 pixels under le mask, our data set corresponds to 76 images, that we can consider as 76 dots in a space of 2754 dimensions

Intuitively one can guess that two identical images will have similar coordinates in the multi-dimensional space. Therefore in the multidimensional space they correspond to two dots located near each other. Conversely, two dissimilar images will correspond to two dots located far away from each other.

Multi-dimensional statistical analysis (MSA), reinforces this idea of "similarity = proximity" but it changes the coordinate system of our data set in order to reduce the number of dimensions to a number a few meaningful axes. These axes or "eigen vectors" correspond to main "trends" or "variations" within our population of images.

1. Absolute values \rightarrow frequencies

 $\mathsf{K}\mathsf{i}\mathsf{j} \to \mathsf{K}\mathsf{i}\mathsf{j} / \Sigma \mathsf{k}\mathsf{i}\mathsf{j} = \mathsf{f}\mathsf{i}\mathsf{j}$

To get the "flavour" of this method

2. Euclidian distance $\rightarrow \chi^2$ distance

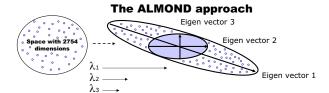
 $fij \rightarrow fij / fi. f.j$

3. Image mass i = fi.

Origine changed to the center of gravity of the table = -f.j

4. Diagonalization of the covariance matrix Xij = (fij - fi. f.j) / fi. f.j

equivalent to a least square fit to define new factorial axes (eigen vectors) and the coordinates of each image on these axes.



One method of diagonalization of the co-variance matrix (T = X' X), called "la méthode de la dragée" or the "Almond

method' illustrates what happens at this stage.

The original multi-dimensional space has been distorted by the chi square matrix to express the variations among the images. Schematically one can say that the cloud of 76 dots (representing our 76 images) which was originally contained in a multi-dimensional "sphere" is now contained within a multi-dimensional "almond".

1. The longest dimension of the almond corresponds to the major "trend" or variation among the image set and is

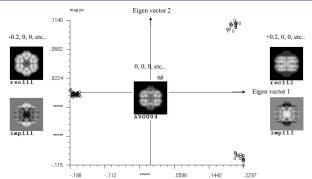
- defined as the first eigen vector. Its amplitude (length) corresponds to the first eigen value \(\lambda 1\). Coordinates of our
- defined as the first eigen vector. Its amplitude (tengin) corresponds to the first eigen value 2.1. Coordinates of our 76 dots along this new axis are calculated.

 Then, the second longest dimension of the almond, orthogonal to the first eigen vector is determined (width of the almond). This second direction corresponds to eigen vector number two and corresponds to the second variation among the images. The amplitude of this second vector is the second eigen value 2.2. Coordinates of our 76 dots along this new axis are calculated.
- Then the third longest dimension of the alond, orthogonal to the first and second eigen vectors is determined (thickness of the almond). This third direction corresponds to eigen vector number three and corresponds to the third variation among the images. The amplitude of this third vector is the third eigen value $\lambda 3$. Coordinates of our 76 dots along this new axis are calculated.

The 76 dots can be projected on planes defines by two selected eigen vectors. Here again the "proximity = similarity" rule applies, and we can identify four types of images in the example set of images.

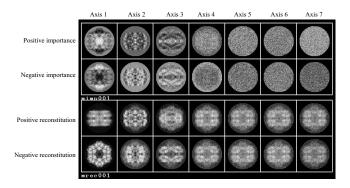
In fact, the information contained in our data is so much compressed that a set of coordinates on the eigen vectors can characterize a given image. Jean-Pierre Brétaudière and Joachim Frank designed "reconstitution and importance images" to express this relationship and to explore the variation related to each eigen vectors.

For example, according to you, how looks an image having for coordinates zero on all eigen vectors?

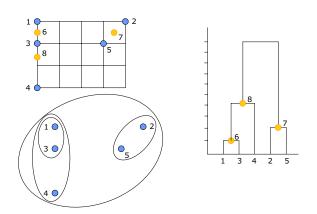


etc.

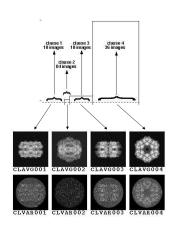
Brétaudière JP and Frank J (1986) Reconstitution of molecule images analyzed by correspondence analysis: A tool for structural interpretation. *J. Microsc.* **144**, 1-14.

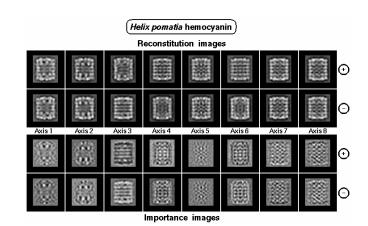


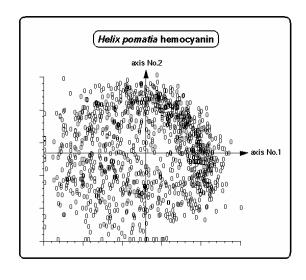
Classification Ascendante Hiérarchique

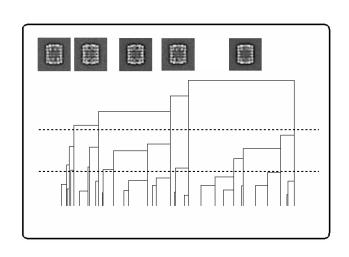


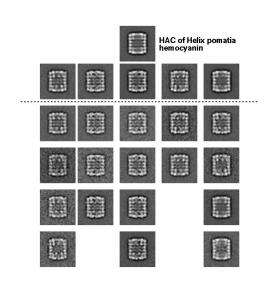
Hierarchical ascendant classification











Hémocyanine of Helix pomatia

External surface rendering

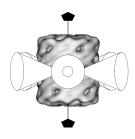
Central cavity

 $\label{eq:Question:Which EM views would you use to suppress the missing cone \ while enforcing a D_5 symmetry?$

If 0° tilt images = TOP views, the missing cone axis is parallel to the five-fold axis of the cylinder.

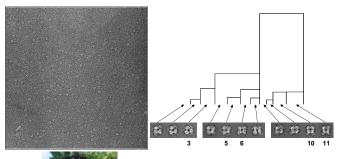
If 0° tilt images = SIDE views the missing cone axis is orthogonal to the five-fold axis of the cylinder.





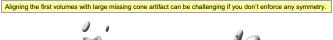
When enforcing symmetry D5 or C5, you will fill up the missing cone of the SIDE views, but the missing cone of the TOP views will always superpose to itself and remain empty.

How to merge volumes when you don't know if your structure has symmetries?



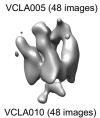


Poster of Magali Cottevieille on the Glutamate synthase complexe. magali.cottevieille@impmc.jussieu.fr



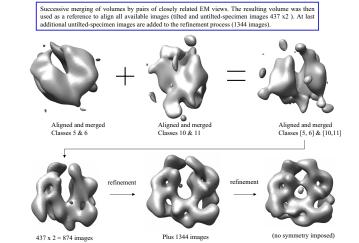


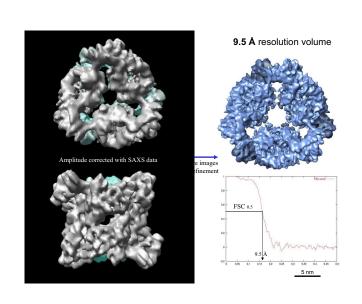






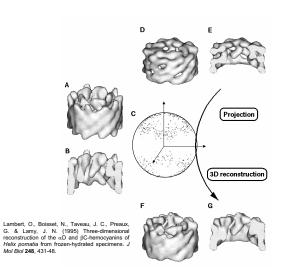
VCLA011 (63 images)

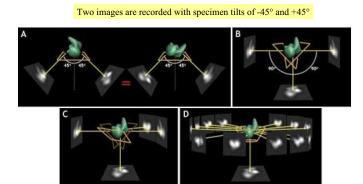




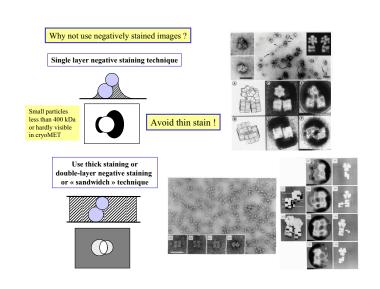
Lambert, O., Boisset, N., Taveau, J. C. & Lamy, J. N. (1994) Three-dimensional reconstruction from a frozen-hydrated specimen of the chiton *Lepidochiton sp.* hemocyanin. *J Mol Biol* **244**, 640-7.

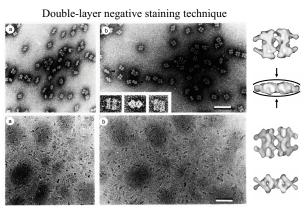
The Orthogonal tilt reconstruction method Andres E. Leschziner & Eva Nogales



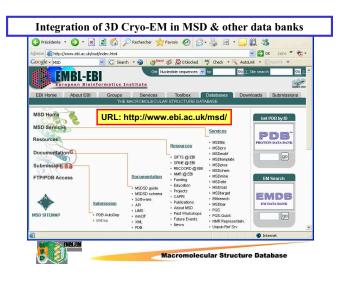


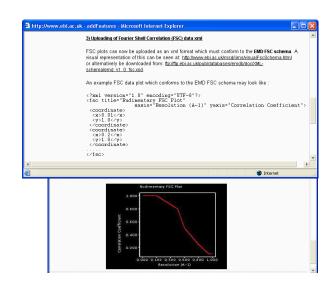
In this case you don't get a conical tilt but the equivalent of a 360° tomogram.



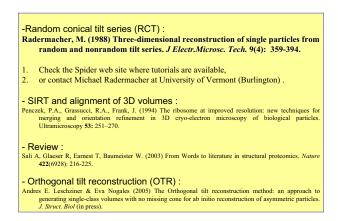


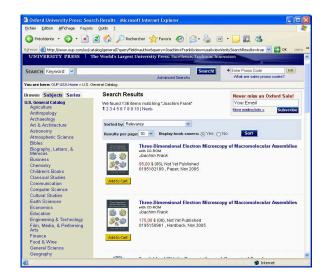
Frozen-hydrated sample





references:





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