

# Improving cryo-EM maps: focused 2D & 3D classifications and focused refinements

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# **Conformational changes of cats?**





# I. Particle sorting



# How to sort out heterogeneity (composition / conformation)? → particle sorting

**<u>4 different approaches in the cryo-EM field:</u>** 

- 1) reference-based, i.e. cross correlation with forward-projections of known structures
- 2) multivariate statistical analysis (MSA): 2D classification or 3D classification, variance analysis + resampling, bootstrapping, 3D resampling
- 3) maximum likelihood based classifications
- 4) deep learning methods



Determining structures of multiple conformational states in a single sample

1) reference-based, i.e. cross correlation with forward-projections of known structures





Loerke et al., Meth. Enzymol. 2010

# Determining structures of multiple conformational states in a single sample 2) multivariate statistical analysis (MSA): 2D classification, 3D classification



distinguish: orientational classification and conformational classification



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Determining structures of multiple conformational states in a single sample

2) multivariate statistical analysis (MSA): 2D classification, 3D classification



**Perform 2 classifications:** 

(i) global MSA for classification according to <u>particle orientations</u> (i.e. classical class averages)
(ii) local MSA <u>with a smaller mask</u> for classification according to <u>particle variability</u>.



Klaholz et al., Nature 2004; see Suppl. Mat.



# Sorting out heterogeneity of complexes:



see also work by P. Penczek (bootstrapping (re-sampling), used primarily to find region of variance) see also S. Scheres/J-M Carazo (maximum likelihood parameter refinement and classification) Simonetti *et al.*, *Nature*, 2008. used by Fischer *et al.*, *Nature*, 2010; Papai *et al.*, *Nature* 2010.

## Determining structures of multiple conformational states in a single sample

3) maximum likelihood based classifications

→ assign particles to different 3D classes based on maximum likelihood
 (probability distribution; uses randomly selected references + ML-weighting)
 Practically:

random subsets are optimized and a low-resolution average structure is used as reference, i.e. <u>resampling</u> is used in combination with likelihood optimization

e.g. Scheres et al., JMB 2005; Meth. Enzymol. 2010;

Lyumkis et al., JSB 2013

Introduction of the ML concept in cryo-EM: F. Sigworth, JSB 1998;

in X-ray crystallography: G. Bricogne, Acta Cryst A, 1991



### **Examples of ML-based 3D classification**



Strong heterogeneity of a reconstituted eukaryotic translation initiation (eIF5B) complex: sorting → 5143 particles, representing 3% of the population in the sample, 6.6 Å reconstruction. Fernández *et al.*, *Science* 2013; V. Ramakrishnan & S. Scheres.



### e.g. ML-based focused classification of 80S / TSV IRES complex with eEF2/GDP/sordarin



#### Abeyrathne et al., eLife 2016



See also: von Loeffelholz et al., Curr. Opin. Struct. Biol. 2017.



e.g. ML-based focused classification

sorting scheme for human 80S/antibiotic complex



Myasnikov et al., Nat. Comm. 2016.



**II. Focused refinement** 



## Local MSA / focused 2D/3D classification & focused refinement:



focused classification



#### see also:

Klaholz et al., Nature 2004; White *et al.*, *JSB* 2004; Penczek et al., JSB 2006; Wong *et al.*, *Elife* 2014;

Concept of focused cryo-EM structure refinement through

- 3D resampling & 3D classification (3D-SC) / bootstrapping
- maximum likelihood 3D classification

using spherical mask or dilated, binarized map of region of interest

Helps: use a slightly larger region than the region of interest, e.g. 30-50 Å in diameter

von Loeffelholz et al., Curr. Opin. Struct. Biol. 2017.



. . .

#### **Focused refinement:**





60S and 40S ribosomal subunits

60S subunit, 40S body and 40S head regions

Natchiar et al., Nature 2017. See also: von Loeffelholz et al., Curr. Opin. Struct. Biol. 2017.







Natchiar et al., Nature 2017.

### Allows visualization of chemical modifications in rRNA:





Natchiar et al., Nature 2017.

**III. Improving data collection quality** 



## **Detectors, dose weighting, movie alignment**



Orlov et al., Biology of the Cell, 2017.



## Volta phase plate data collection facilitates image processing and cryo-EM structure determination



dfVPP data behave more robustly during image processing: particle selection, accuracy in alignments, 2D & 3D classifications, map interpretation



MSA-based classification (3 particles)

von Loeffelholz et al., JSB 2018.



Single- / dual-tilt cryo electron tomography



*Titan Krios, GIF/K2, VPP, SerialEM, Tom toolbox, IMOD* **T. Frosio & J. Ortiz / Klaholz lab.** 



see also Myasnikov et al., Ultramicroscopy 2013.

### **Summary**

**Improving 3D reconstructions and cryo-EM maps:** 

- particle sorting (to address heterogeneity)
   focused refinement (→ composite maps; for a given conformation)
   improve data quality (CMOS detectors, VPP, dual-axis tomography)
- 4) to help interpretation: map sharpening
  - (B-factor, bp-filtering, LocScale, phenix.autosharpen etc.)

# *Keep in mind:*

- cryo-EM maps are electrostatic potential maps ( $\rightarrow$  Glu's, Asp's etc.)
- initial 3D reconstruction (not "model")
- $\rightarrow$  atomic <u>model</u> building & refinement, proper geometry & B-factors

