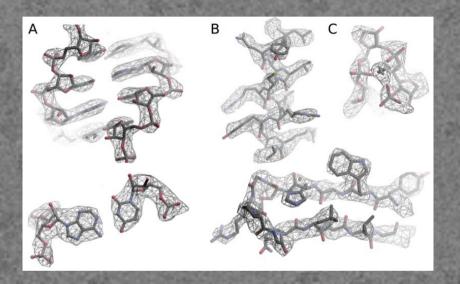
Methods for heterogeneity analysis

(with bias)



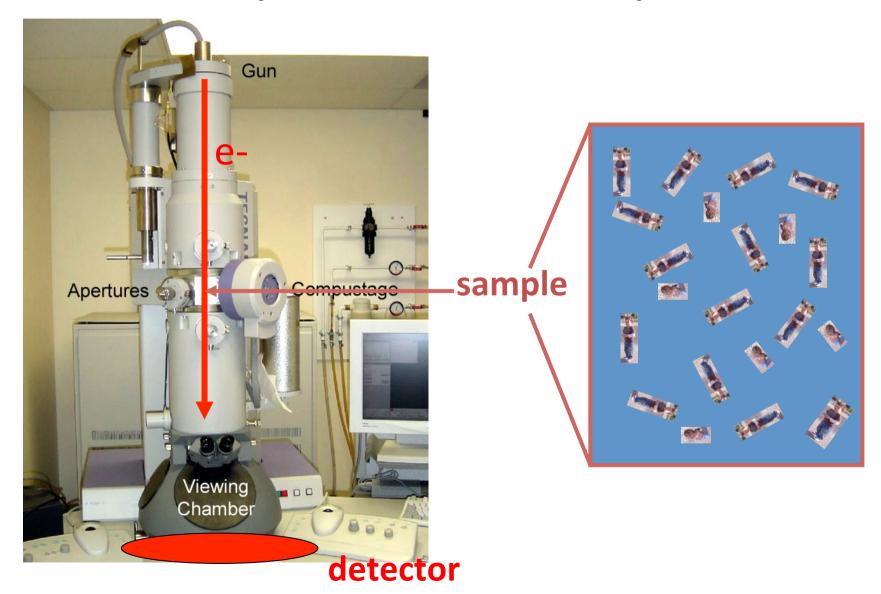
Sjors H.W. Scheres EMAN workshop, Houston, October 2015



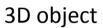
An example "protein"

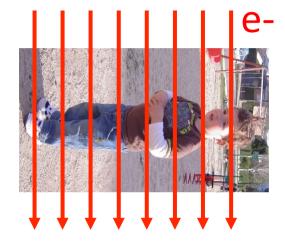


Experimental setup



Electron microscopy imaging





2D projection



We collect data in 2D, but we want 3D info!

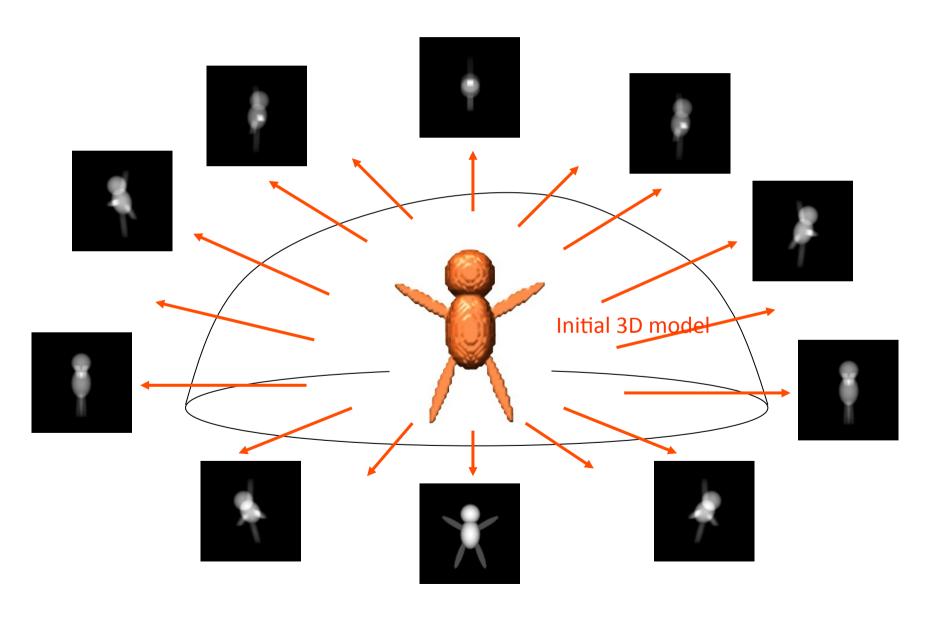
Single particle analysis

• Embedded in ice: many unknown orientations

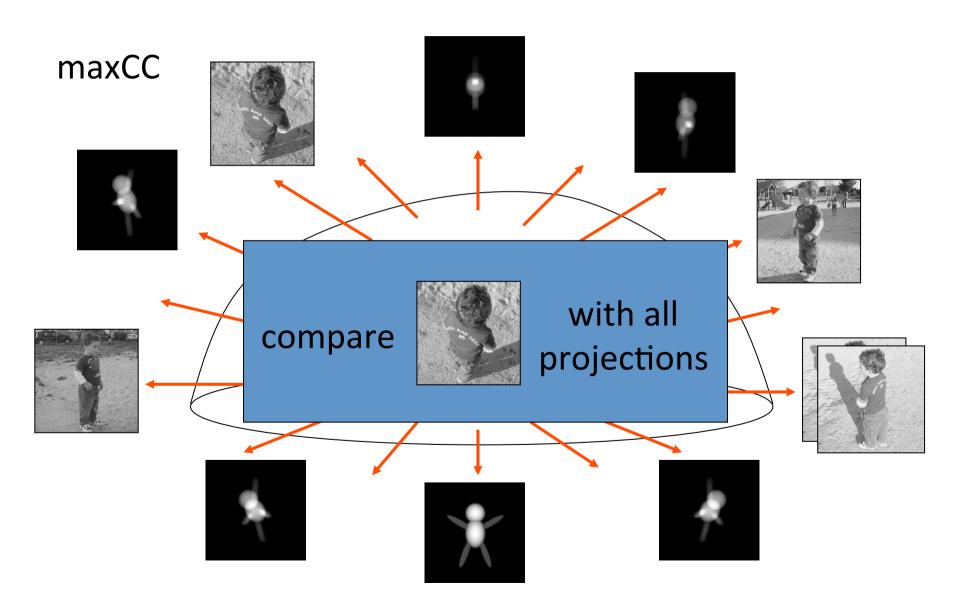


Combine all 2D projections into a 3D reconstruction

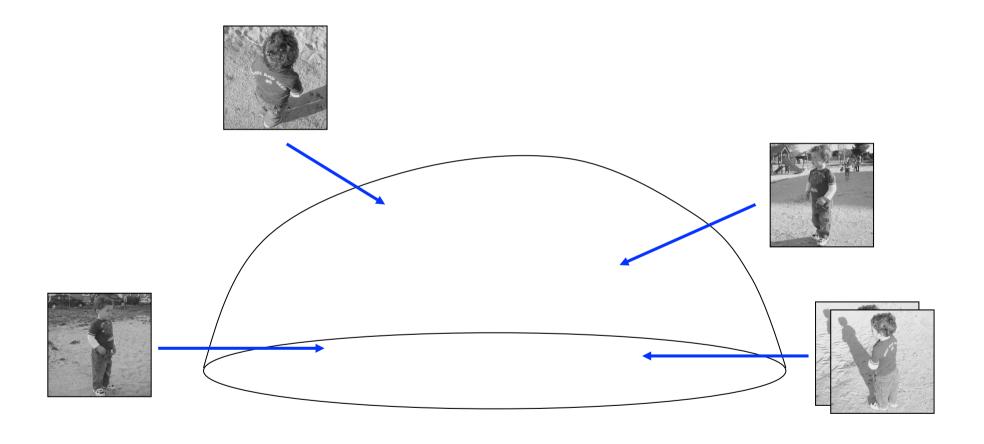
Projection matching



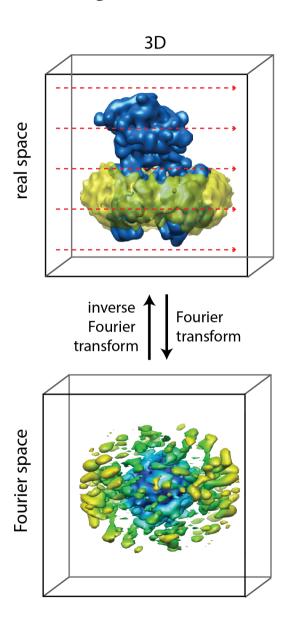
Projection matching



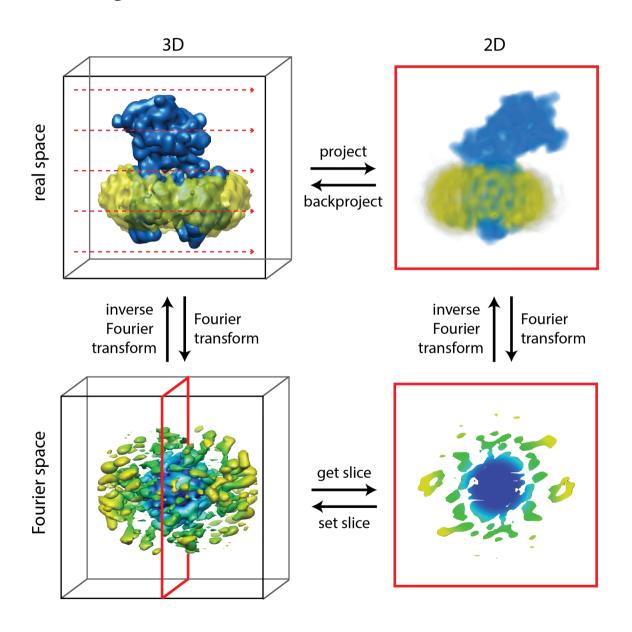
3D reconstruction



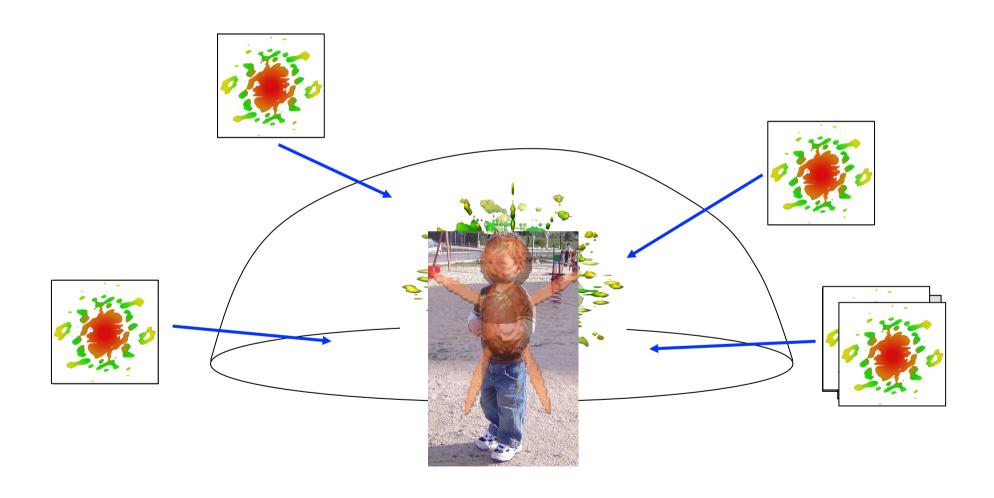
Projection slice theorem



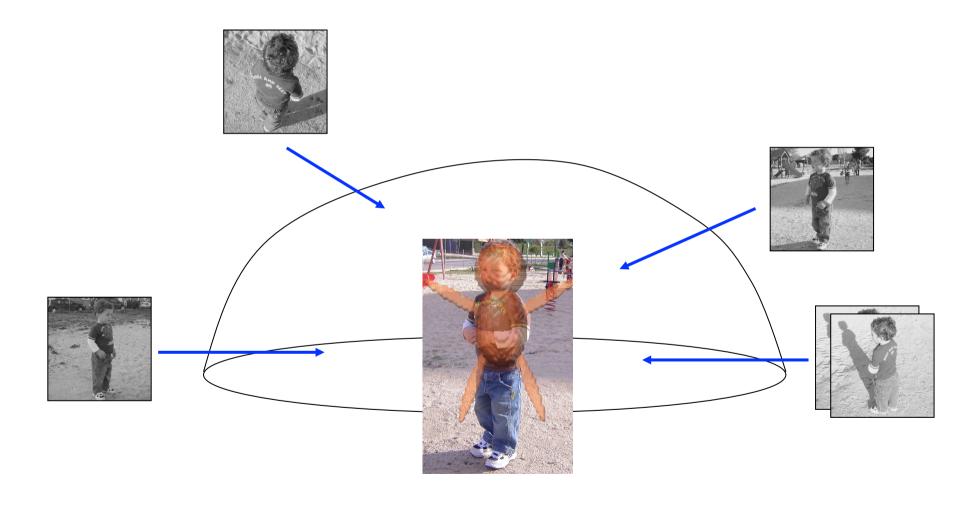
Projection slice theorem



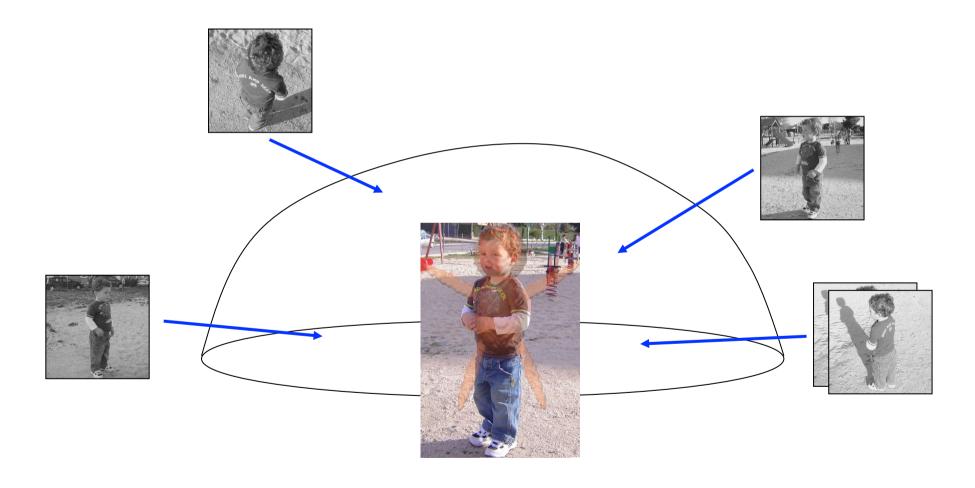
Iterative refinement



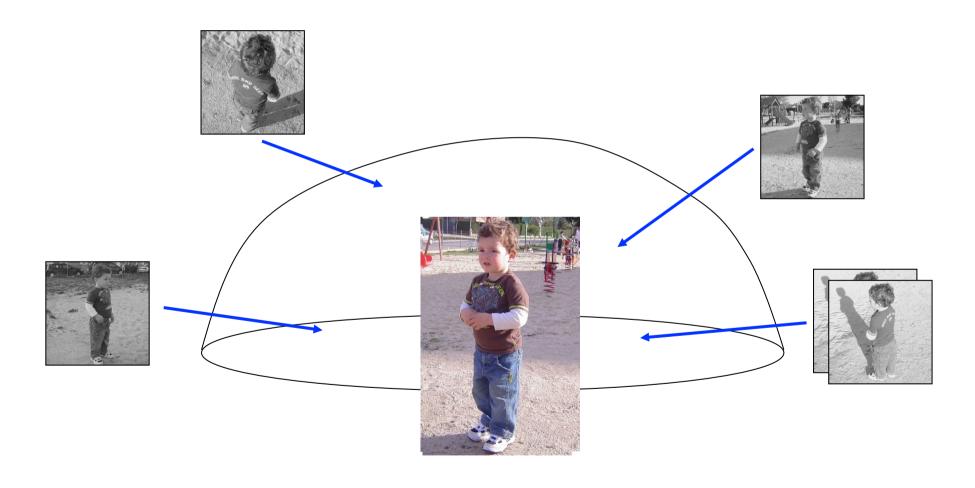
3D reconstruction



Iterative refinement

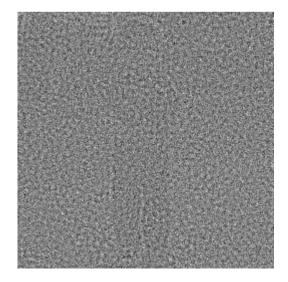


Iterative refinement



Further inconveniences

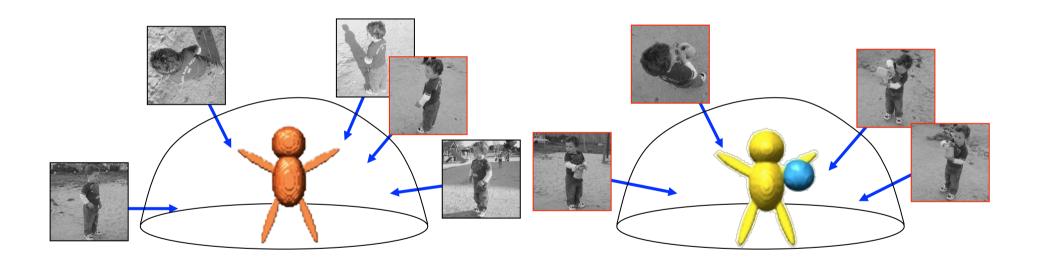
- Defocussing & microscope imperfections introduce artefacts (-> CTF correction)
- Low dose: large amounts of noise
- Structural heterogeneity!



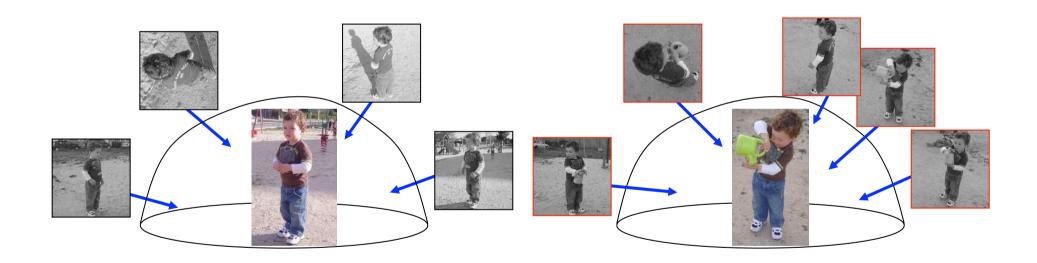
All samples are structurally heterogeneous!



Multi-reference refinement

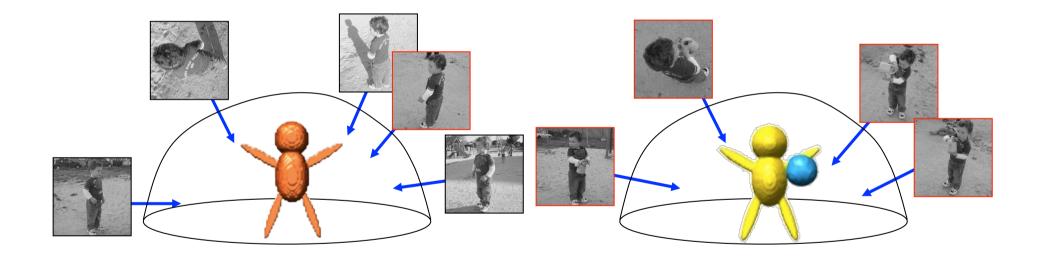


Multi-reference refinement



Supervised classification

(developed in the Frank lab)

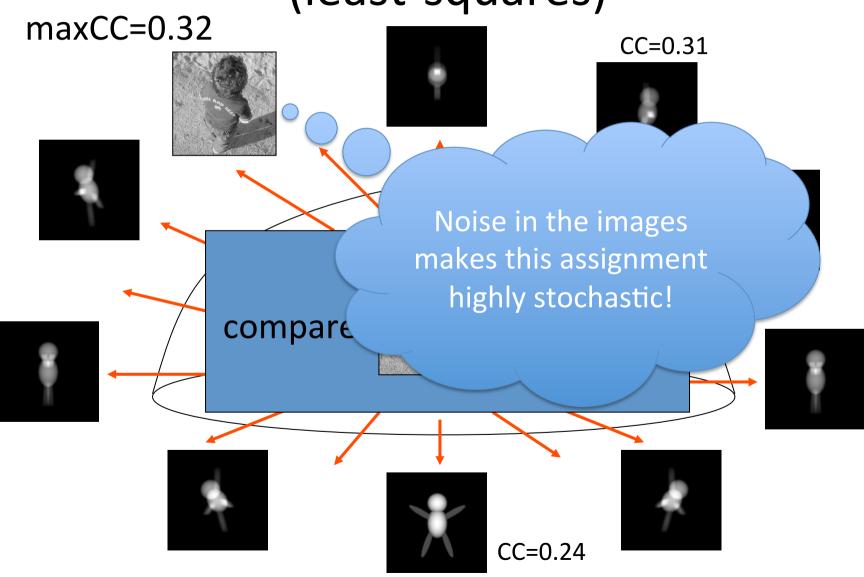


You kind-of need to know the answer already....

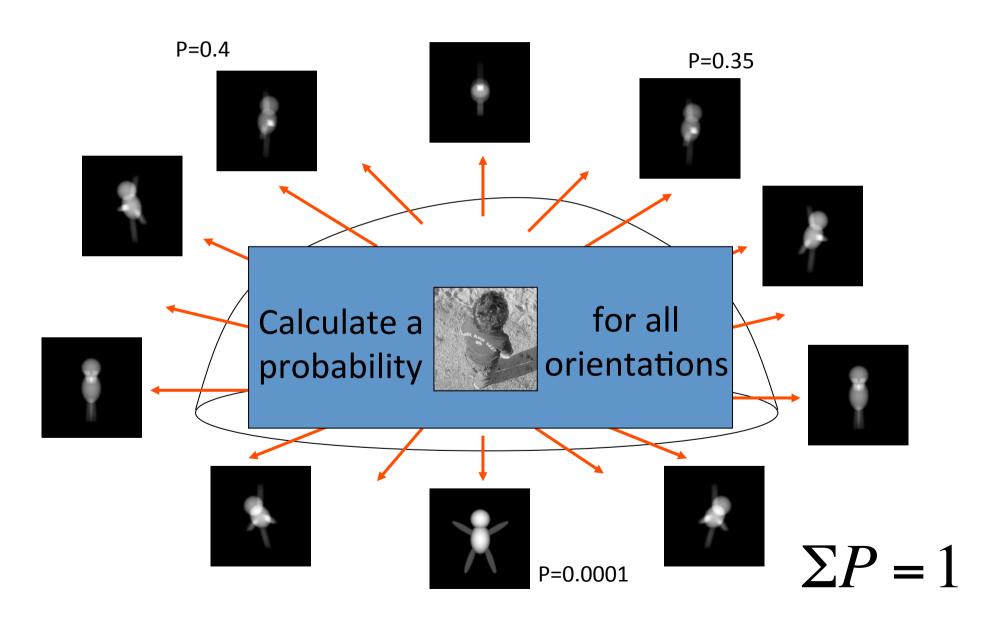
Maximum-likelihood approaches

- Marginalize over orientations & classes
 - Probability-weighted assignments
- First described by Fred Sigworth (JSB, 1998)
 - For 2D-alignment, single-reference
 - Real-space data model (white-noise model)
 - Matlab scripts
- Then extended for 2D & 3D classification (2005-2010)
 - XMIPP Scheres et al, JMB 2005; Nat Methods 2007;
- 3D ML-based classification without marginalizing over orientations
 - FREALIGN
 Lyumkis et al, JSB, 2013

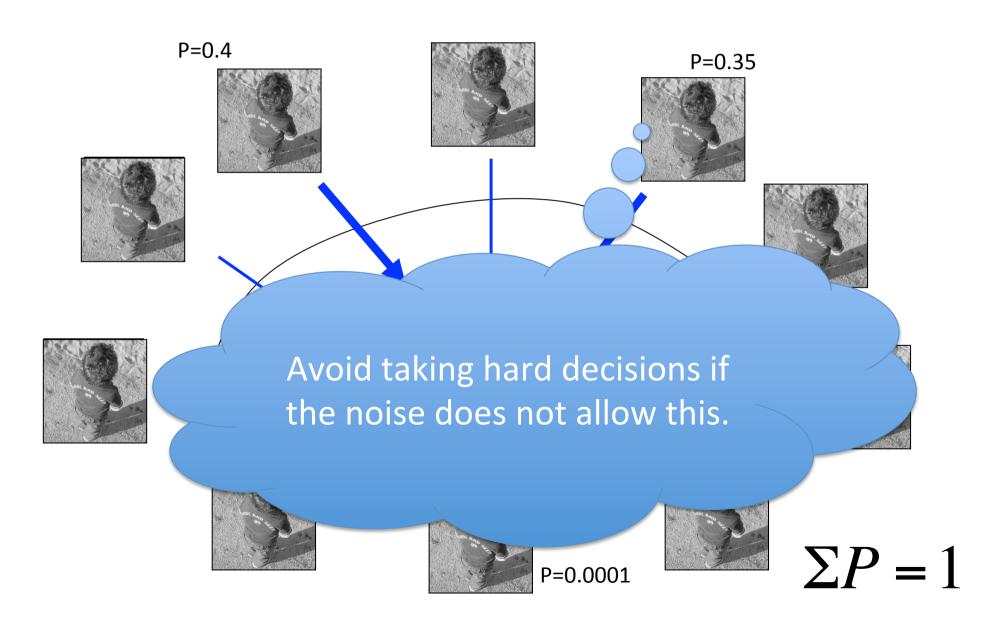
Maximum cross-correlation (least-squares)



Maximum likelihood



Maximum likelihood

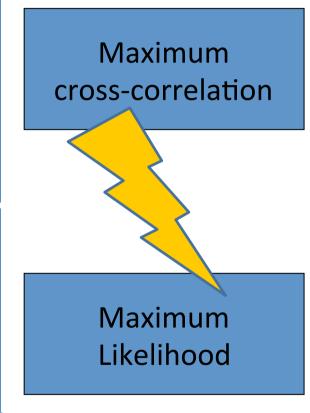


Incomplete data problems

Option 1: add Y to the model

In the limit of **noiseless data** the Two techniques are equivalent!

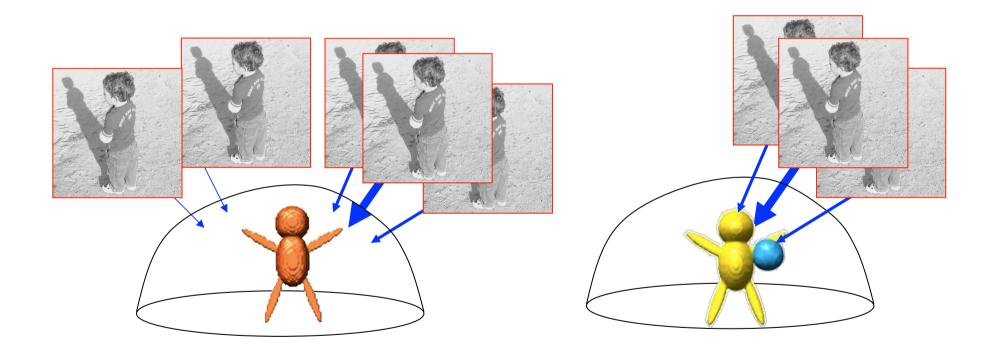
Option 2: marginalize over Y



$$L(\Theta) = P(X \mid \Theta) = \int_{Y} P(X \mid Y, \Theta) P(Y \mid \Theta) d\phi$$

Probability of X, regardless Y

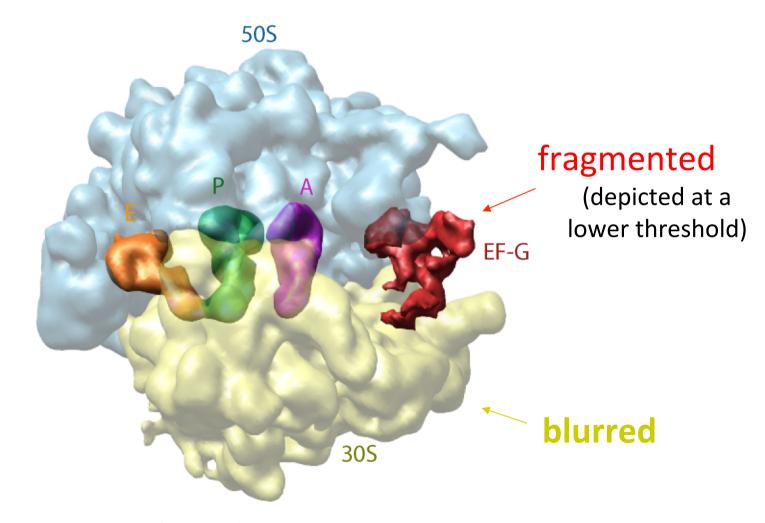
ML3D classification



Probability-weighted angular & class assignments

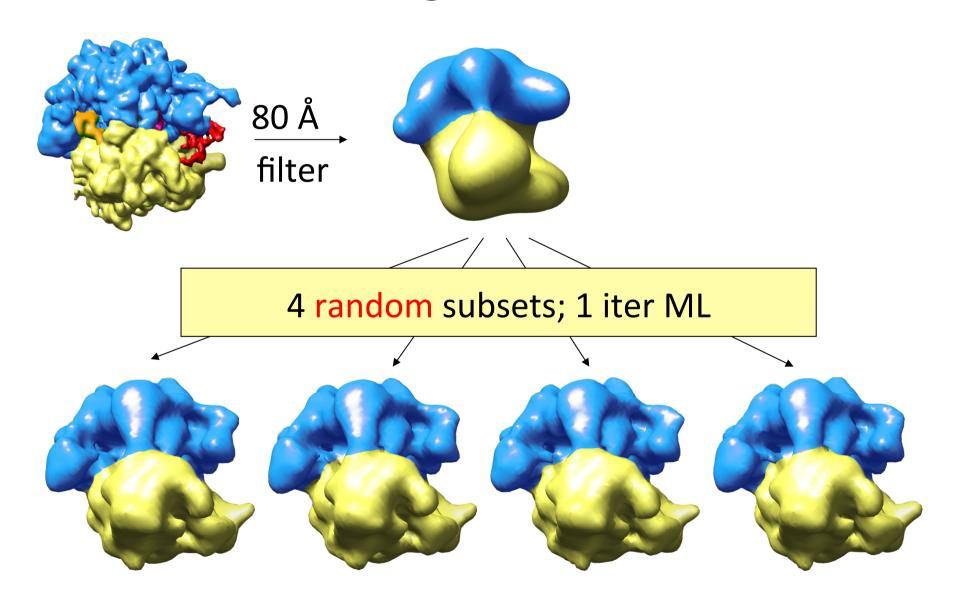
Prelim. ribosome reconstruction

91,114 particles; 9.9 Å resolution



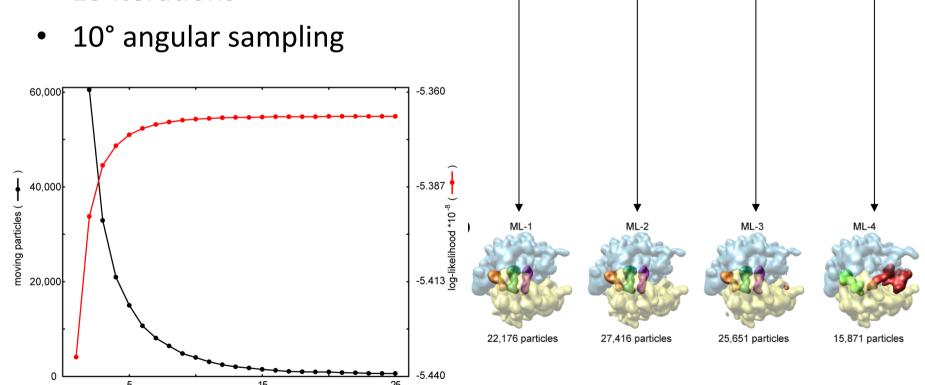
In collaboration with Haixiao Gao & Joachim Frank

Seed generation



ML3D-classification

- 4 references
- 91,114 particles
- 64x64 pix (6.2Å/pix)
- 25 iterations



Scheres et al, Nat Methods, 2007

Regularised likelihood approach

- Data model in Fourier-space
 - Colored (correlated) noise
 - CTF-correction
- Marginalize over orientations & classes
 - Probability-weighted assignments
- Regularization term
 - Penalize high-frequency components
 - Elegant derivation of 3D Wiener filter
 - Iteratively learn power of signal and noise from the data
 - No user-expertise required to optimally filter data/map
 - Objectivity
- RELION

Scheres, JMB 2012; JSB 2012

Other 3D classification tools (I)

- Non-ML multi-reference refinement
 - IMAGIC/SPIDER Van Heel / Frank labs
 - EMAN2 (new similarity measures, alternate 2D/3D)
 Tang et al, JSB 2012; Ludtke et al, JSB 1999
 - SIMPLE (stochastic hill-climbing)

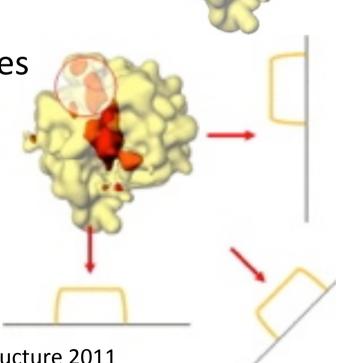
Elmlund&Elmlund, JSB 2012

- Multi-variate statistical analysis
 - IMAGIC/SPIDER

Elad et al, JSB 2008

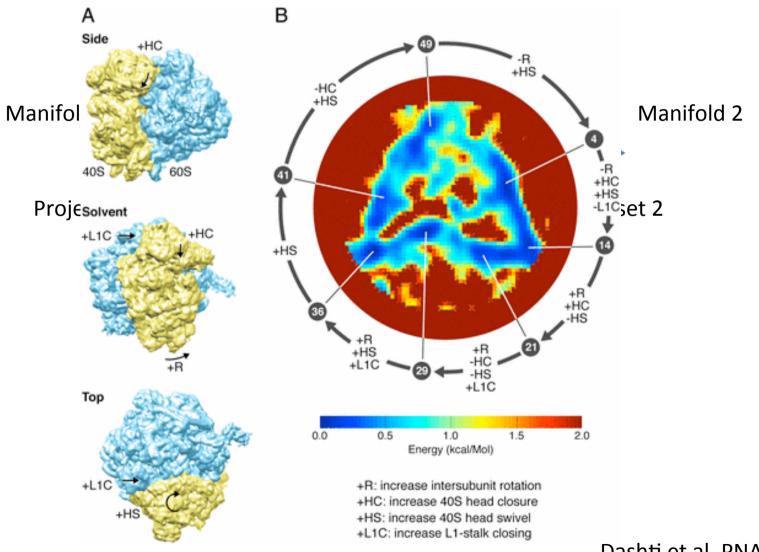
Other 3D classification tools (II)

- Boot-strapping & 3D (co-)variance map
 - Detect and quantify heterogeneity!
- Focused classification
 - Mask out relevant areas in images
- MSA of bootstrapped maps
 - More generally applicable
 - Pawel: SPARX



Classification of a continuum of states, and mapping of the energy landscape

Joachim Frank (Columbia), Peter Schwander and Abbas Ourmazd (U. of Wisconsin)

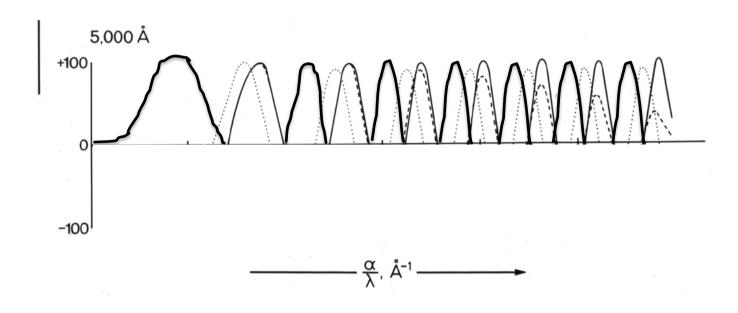




Many variations/applications

Possible in different software packages

Phase flipping



- Easy to do
- Reasonably effective
- Problems in classification?

(3D) Wiener filter

$$V = \frac{\sum_{i=1}^{N} \mathbf{P}_{\varphi}^{T} \frac{\text{CTF}_{i}}{\sigma_{i}^{2}} X_{i}}{\sum_{i=1}^{N} \mathbf{P}_{\varphi}^{T} \frac{\text{CTF}_{i}^{2}}{\sigma_{i}^{2}} + \frac{1}{\tau^{2}}}$$

Optimal linear filter

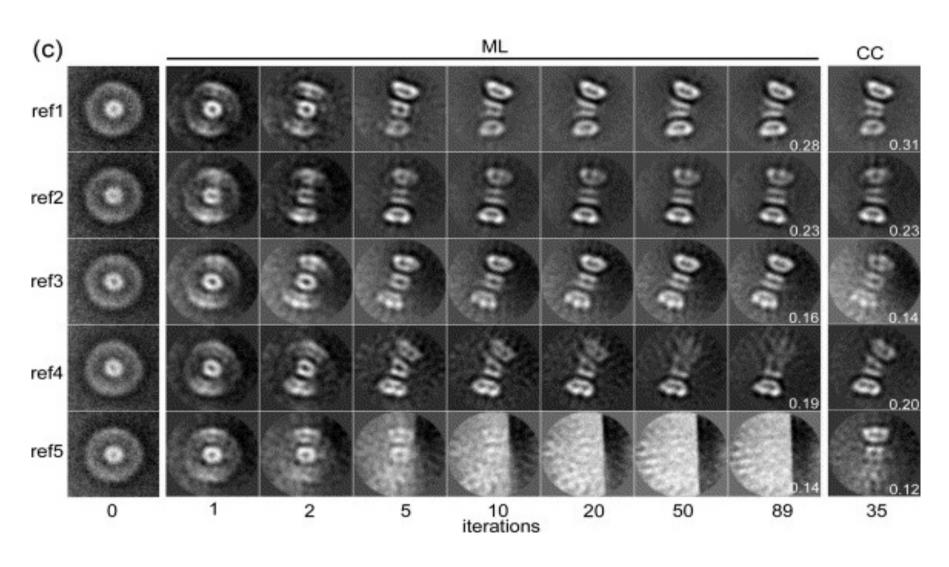
• σ^2 : noise power • τ^2 : signal power

- Low-pass filters & corrects for CTF
- τ^2/σ^2 is often approximated as a constant => low-pass filter effect is lost
- You cannot pre-Wiener filter your data!

2D classification

- Multi-reference 2D refinement/alignment
 - RELION, XMIPP, EMAN2, SPARX (ISAC), SPIDER,
 IMAGIC
- MSA/PCA
 - SPIDER, IMAGIC, XMIPP, EMAN/SPARX?

Reference-free 2D classification



2D classification

- We ALWAYS do 2D class averaging to tidy up the data set
 - Use at least ~100 particles/class for cryo-EM
 - Fewer for negative stain
- Often:
 - Large, high-resolution classes with nice particles
 - Small, low-resolution classes with crap
- Delete bad classes (and possibly repeat)

3D classification

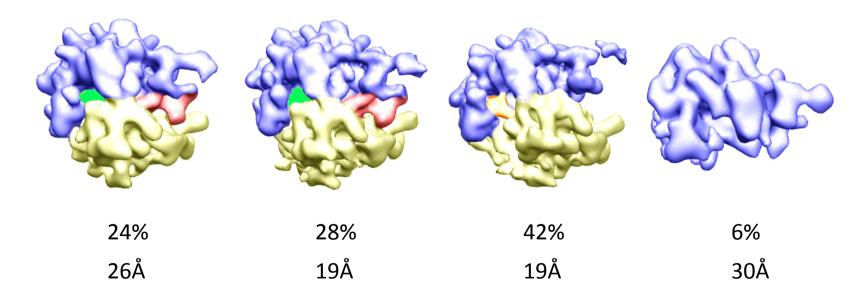
- We ALMOST ALWAYS do 3D classification
 - Almost all samples are heterogeneous!
 - Use at least ~3,000 particles/class for cryo-EM
 - Computational cost often limits to 4-10 classes.

Main scenarios:

- 7.5° angular sampling; exhaustive angular searches
- Finer angular sampling (e.g 0.9° or 1.8°); local searches around angles from 3D single-reference refinement
- NEW: keep angles fixed and only classify (within a mask)
 - good for presence/absence of small factor

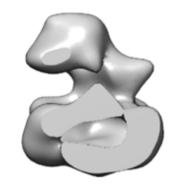
Classify structural variability

- Standard data set from the Frank lab
 - 10,000 70S ribosomes (50% +EFG; 50% -EFG)
 - MAP-refinement K=4



Data cleaning

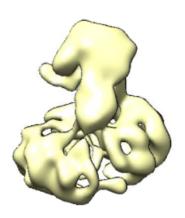
- One/few good classes
- Discard bad classes

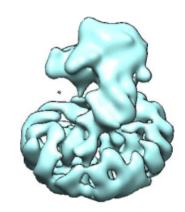


γ-secretase



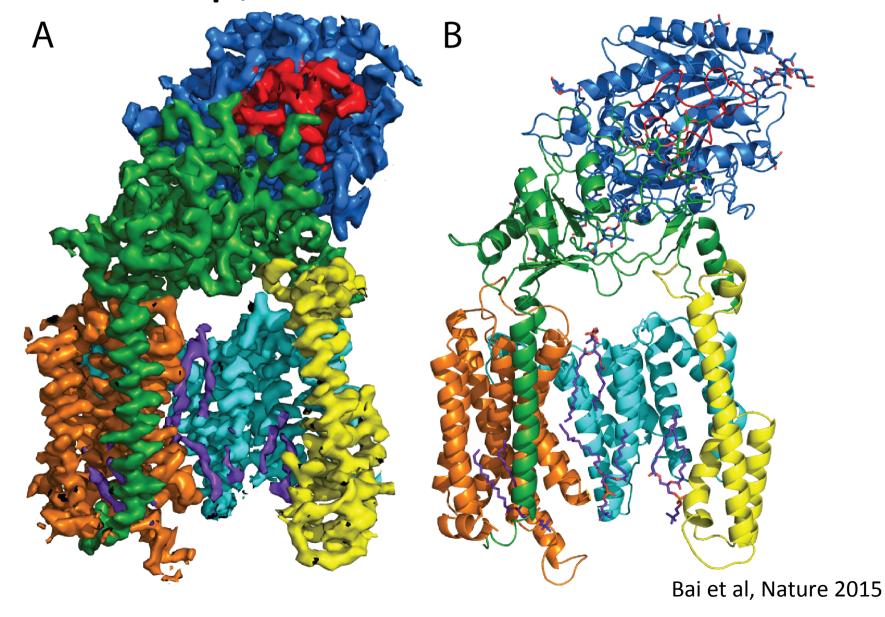


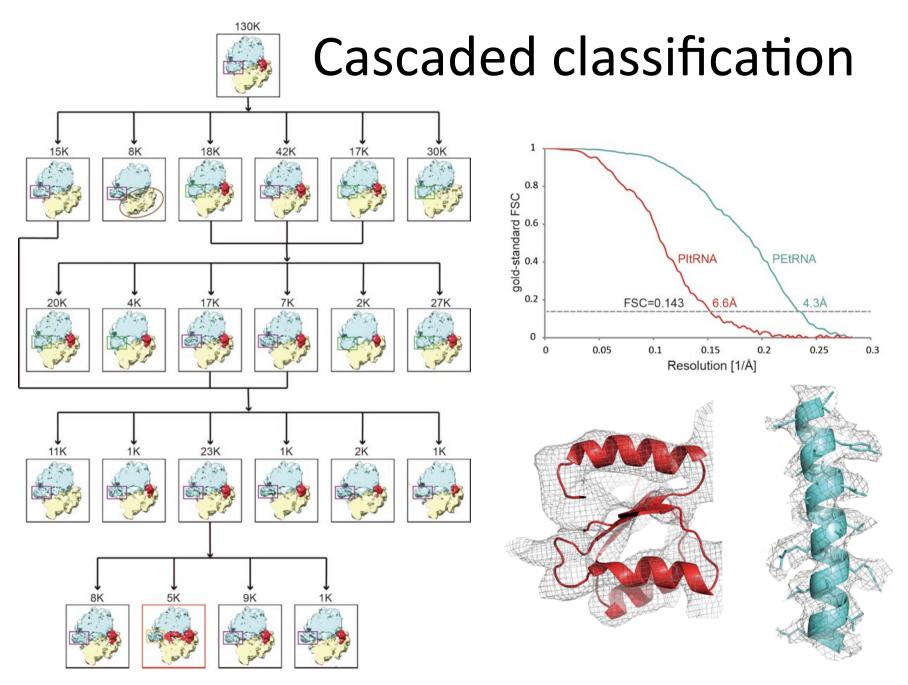




Lu et al, Nature, 2014

3.4 Å map, ~130 kDa ordered mass

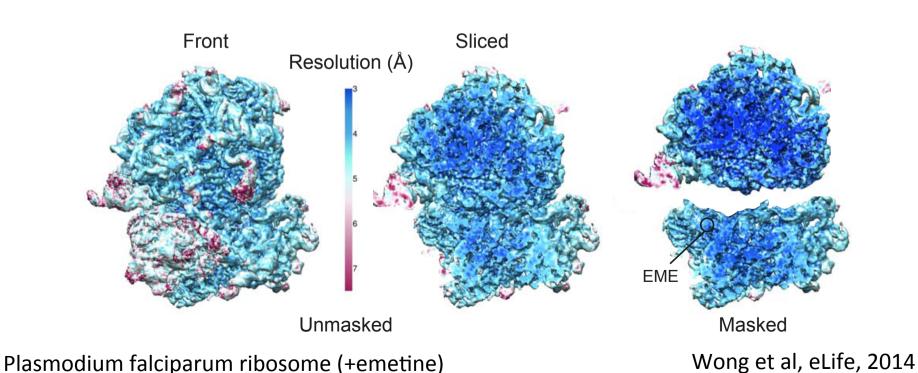




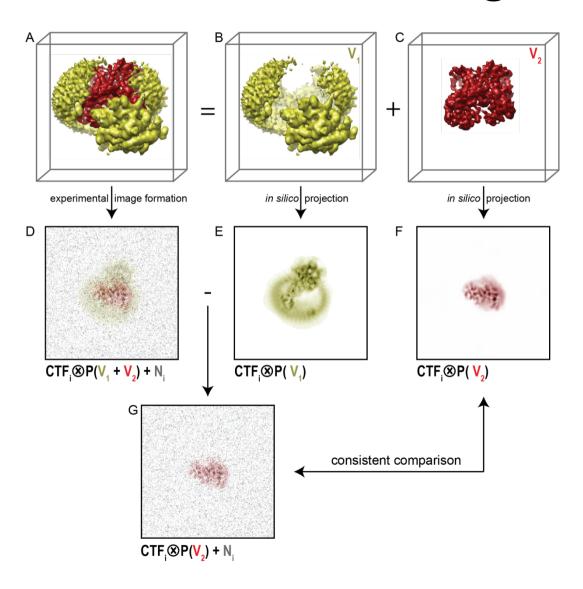
Fernandez et al, Science, 2013

Continuous heterogeneity: Masked refinements

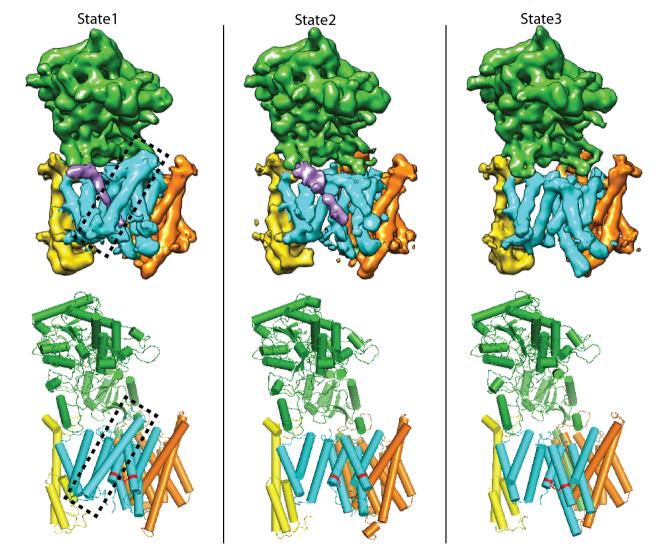
 Mask out volume of interest in reference at every step of 3D-(single-reference) refinement



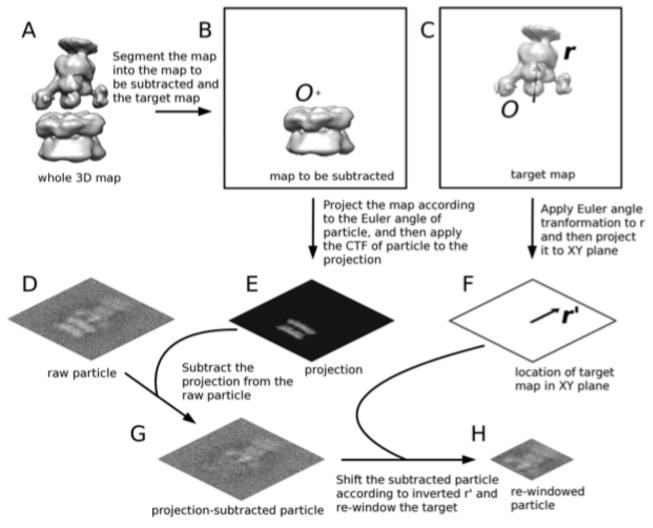
Masked classification + signal subtraction



Conformational heterogeneity

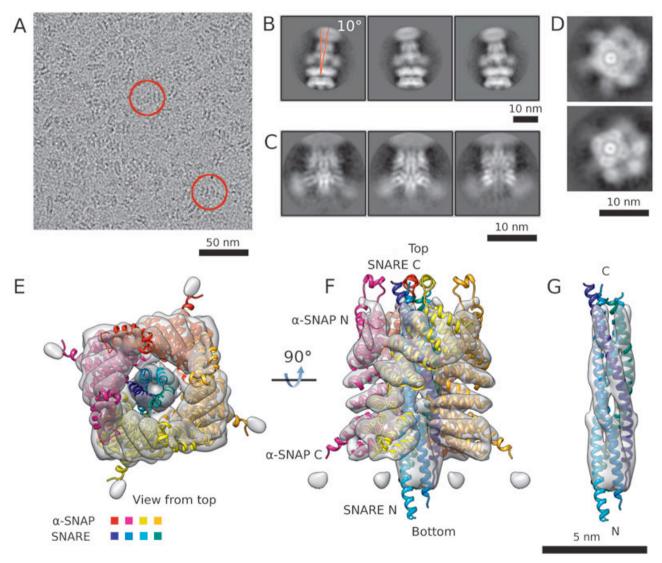


Independent development

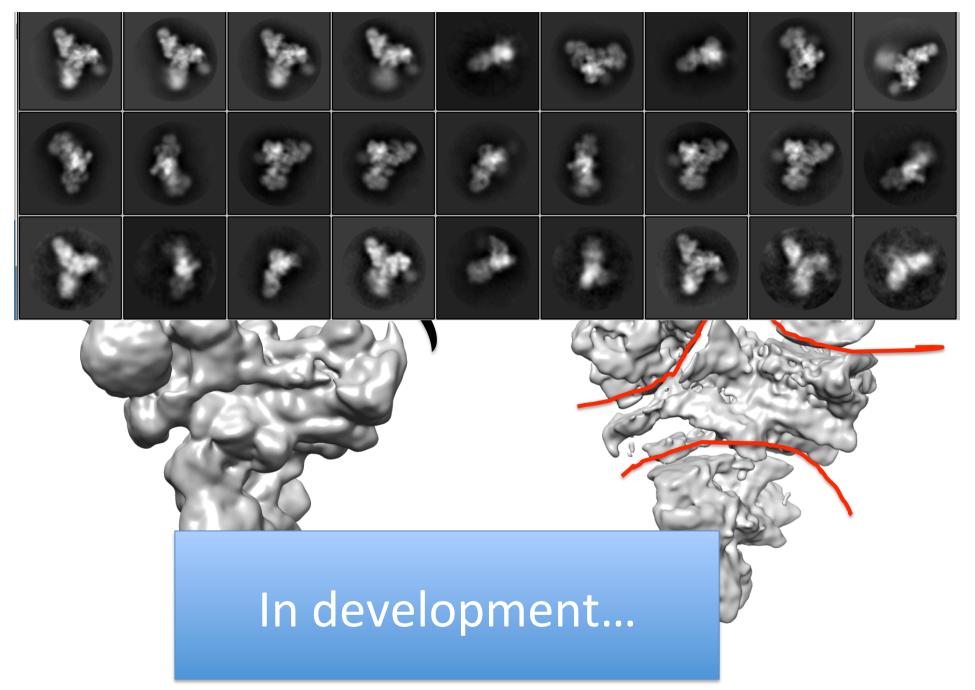


Zhou, ..., Hongwei Wang, Senfang Sui Cell Research, Apr 2015

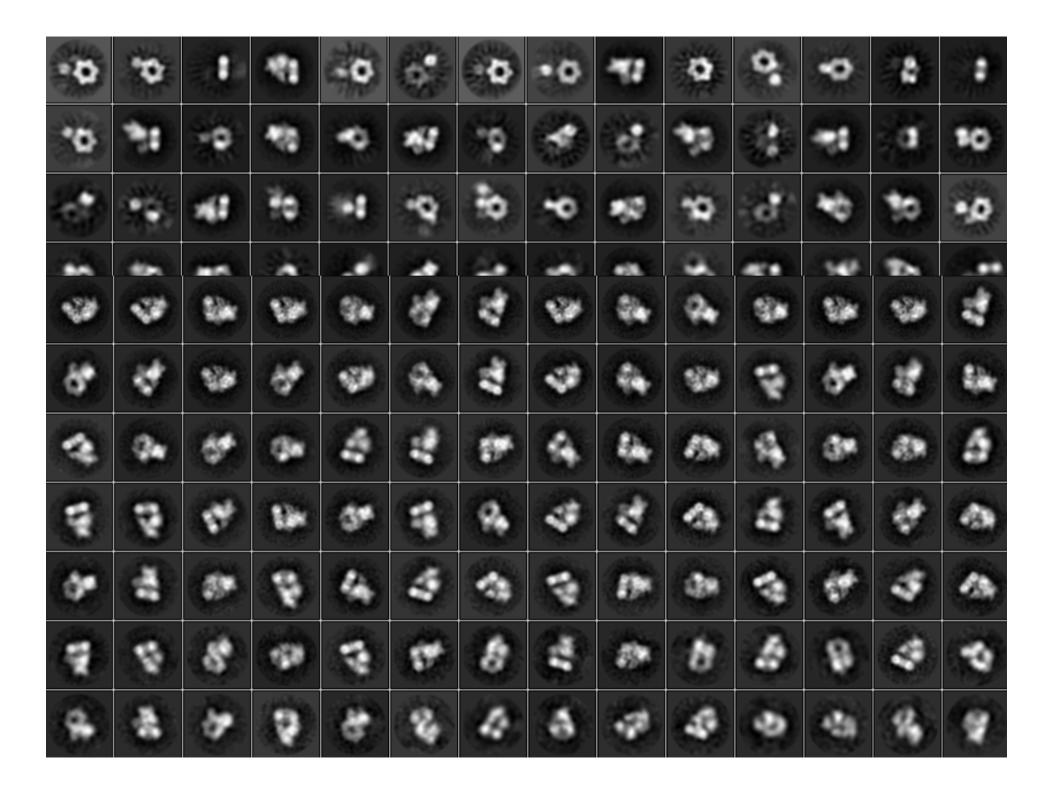
SNAP-SNARE



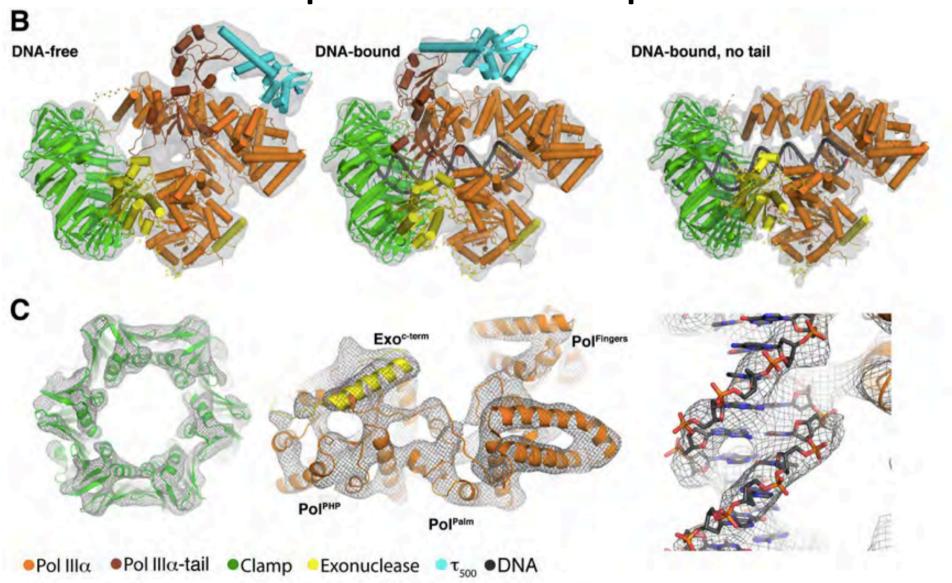
Zhou, ..., Hongwei Wang, Senfang Sui Cell Research, Apr 2015



Some mistakes to avoid...



Replication complex



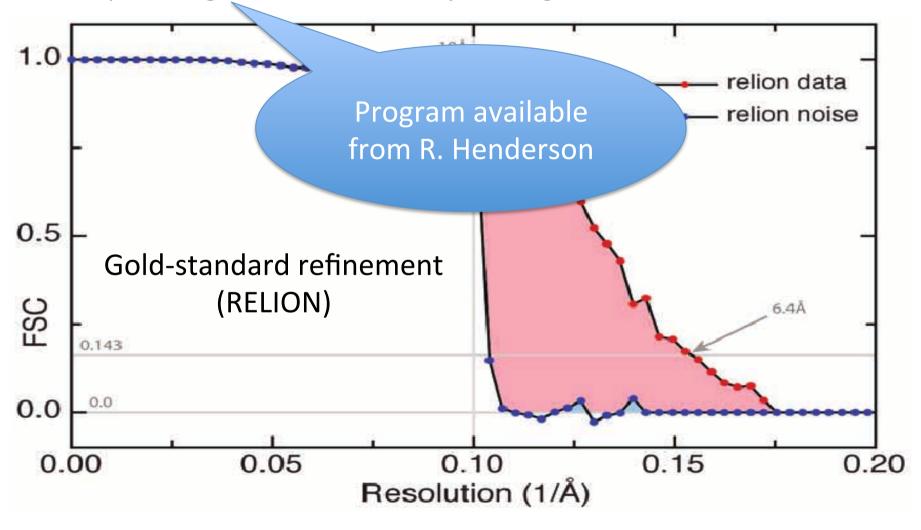
Overfitting

 Always use gold-standard refinement OR limited resolution refinement

- Some new algorithm?
 - Test high-resolution noise substitution

High-resolution noise-substitution

Replace signal in the data beyond a given resolution d with noise



Get stuck with a wrong initial model

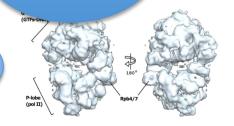
No program is guaranteed to find the global minimum...

Human RNA polymerase II PIC He et al & Nogales, Nature (2013)

Nogales, Nature (2013)

As resolutions improve, this will be ever less of a problem.

Should we stop publishing blobs?







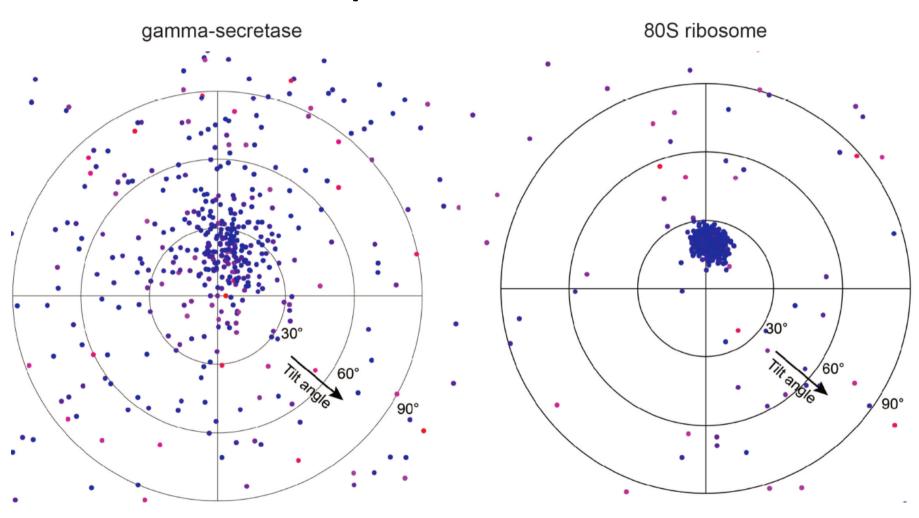






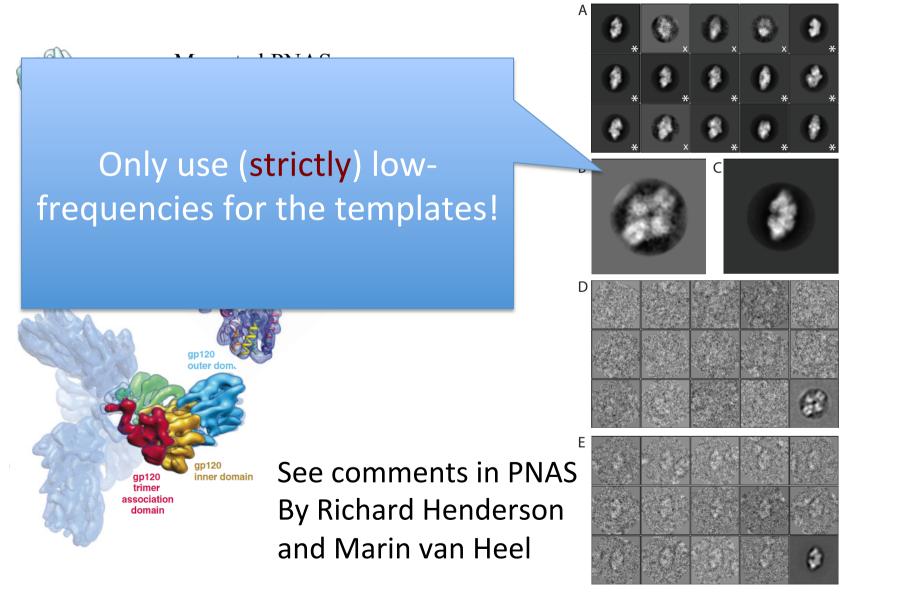


Tilt-pair validation



(like in RELION-1.3)

Template-based auto-picking







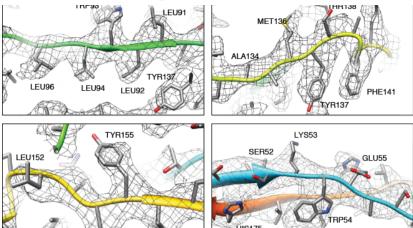
Microscopes: FEI, Jeol, Zeiss, ...

Detectors: K2, Falcon, DE, TVIPS, ...

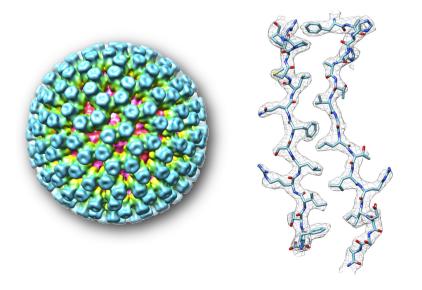
Software: SPIDER, IMAGIC, EMAN, SPARX,

XMIPP, BSOFT, FREALIGN, RELION, ...

Wang et al (2014) Nat Comm.



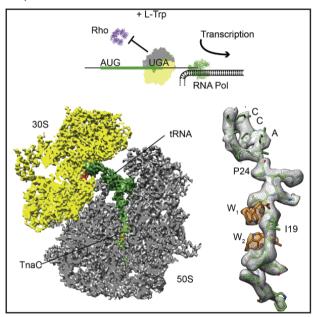
JEOL3200, DE-12, EMAN (3.8 Å)



Cell Reports

Molecular Basis for the Ribosome Functioning as an L-Tryptophan Sensor

Graphical Abstract



Authors

Lukas Bischoff, Otto Berninghausen, Roland Beckmann

Correspondence

beckmann@lmb.uni-muenchen.de

In Brief

Bischoff et al. now present a cryoelectron microscopy reconstruction of a TnaC stalled ribosome, revealing two L-Trp molecules in the ribosomal exit tunnel. As a result, the peptidyl transferase center adopts a distinct conformation that precludes productive accommodation of release factor 2.

Titan Krios, Falcon-II, SPIDER (3.8 Å)

Tim Grant & Niko Grigorieff, eLife 2015

Titan Krios, K2, FREALIGN (2.6 Å)

Conclusions

- Image processing will continue to drive this field forward
 - A variety of software solutions will be most efficient
- New hardware will continue to have huge impacts
 - Better SNRs: distinction between smaller differences
- Making good samples remains crucial!
 - Good classification algorithms are no excuse for bad samples...
- Structural heterogeneity can be an opportunity!
 - If addressed adequately

Thanks!

LMB EM-course 2014

Daily in the MPLT from 9:30-10:30am

Mon May 12: Tony Crowther

Course introduction with a historical perspective

Tue May 13: Sjors Scheres

Image formation, Fourier analysis, CTF theory

Wed May 14: Chris Russo

Microscopy physics and optics

Thu May 15: Lori Passmore

sample preparation

Fri May 16: Paula da Fonseca

Initial data analysis

Mon May 19: Sjors Scheres

Image refinement in 2D and 3D

Tue May 20: Tanmay Bharat

Tomography and sub-tomogram averaging

Wed May 21: Richard Henderson

Map validation

Thu May 22: David Barford & Alan Brown

Low- and high-resolution modeling

Thu May 22: Shaoxia Chen, Christos Savva & others

(11am-12pm) Local setup and training & 2 example applications

Enquiries: scheres@mrc-lmb.cam.ac.uk

Lecture PDFs and professionally edited videos available on:

ftp://ftp.mrc-lmb.cam.ac.uk/pub/scheres/EM-course