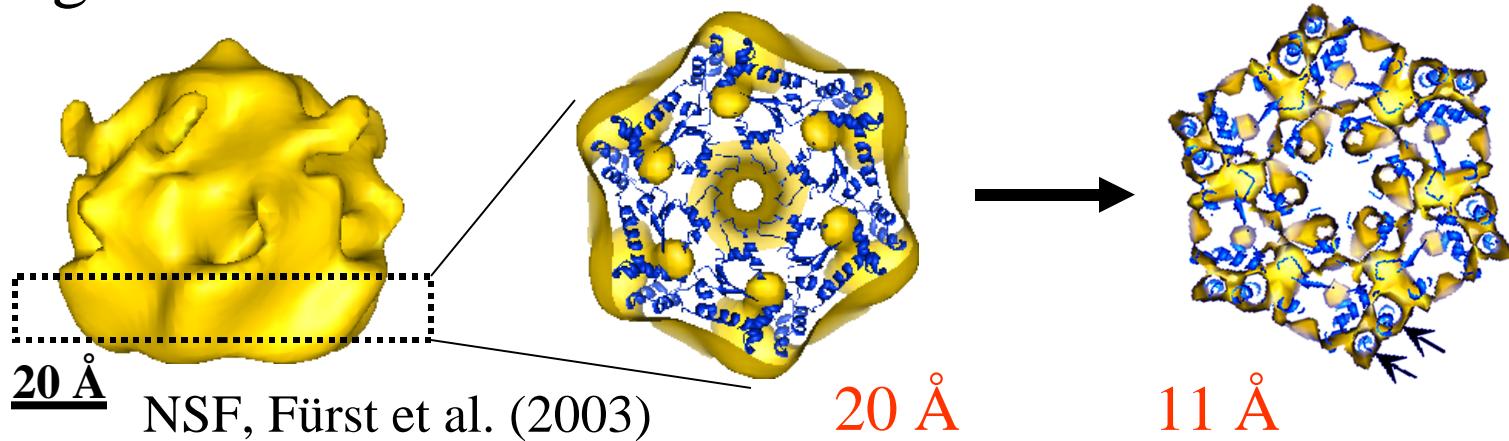


Refinement Strategies for Single Particle Structure Determination

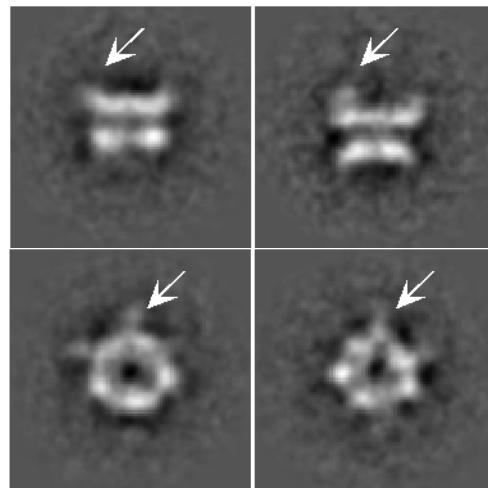
N. Grigorieff

Goals

- Higher resolution



- Sorting of structural heterogeneity

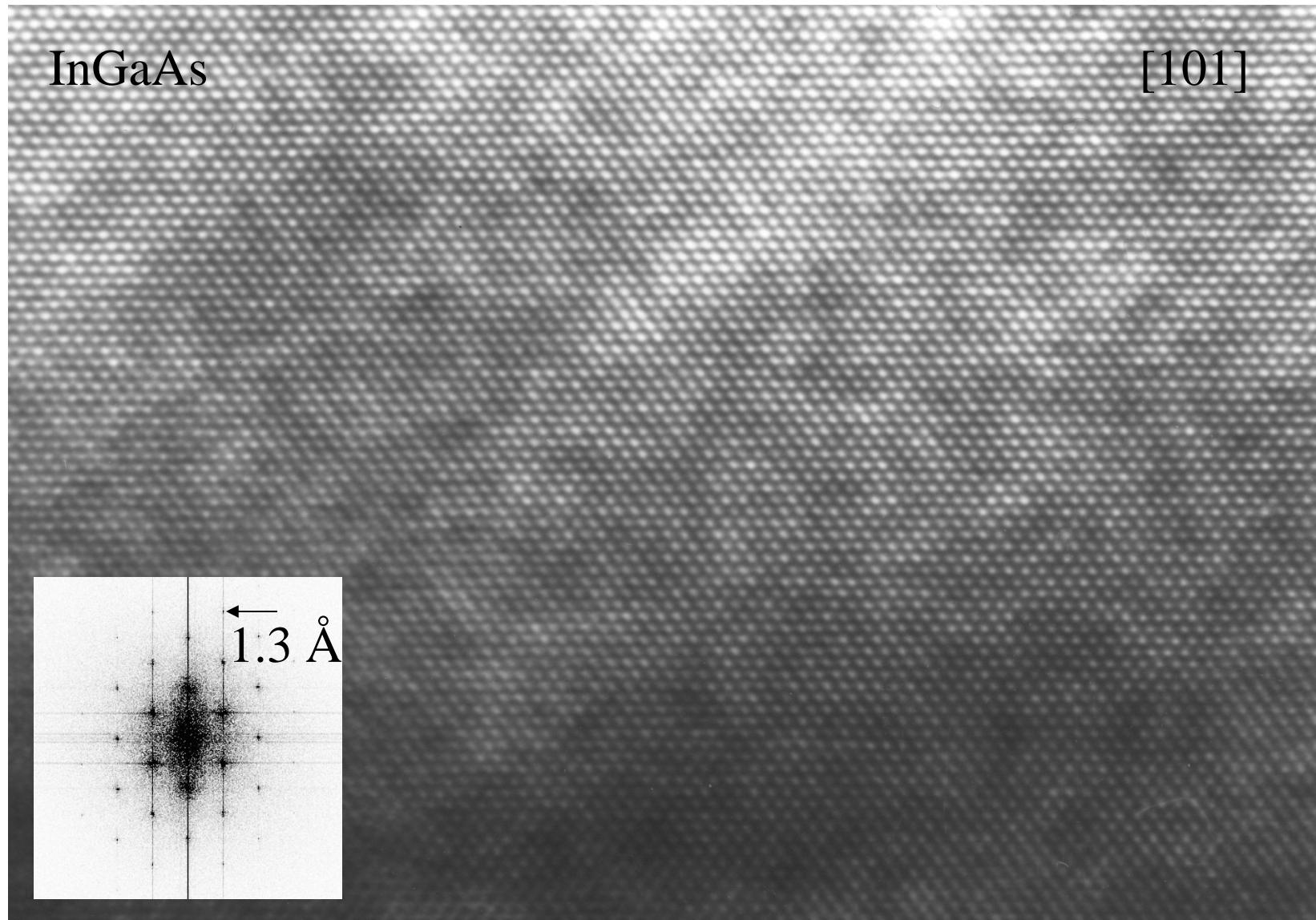


The Prophecy

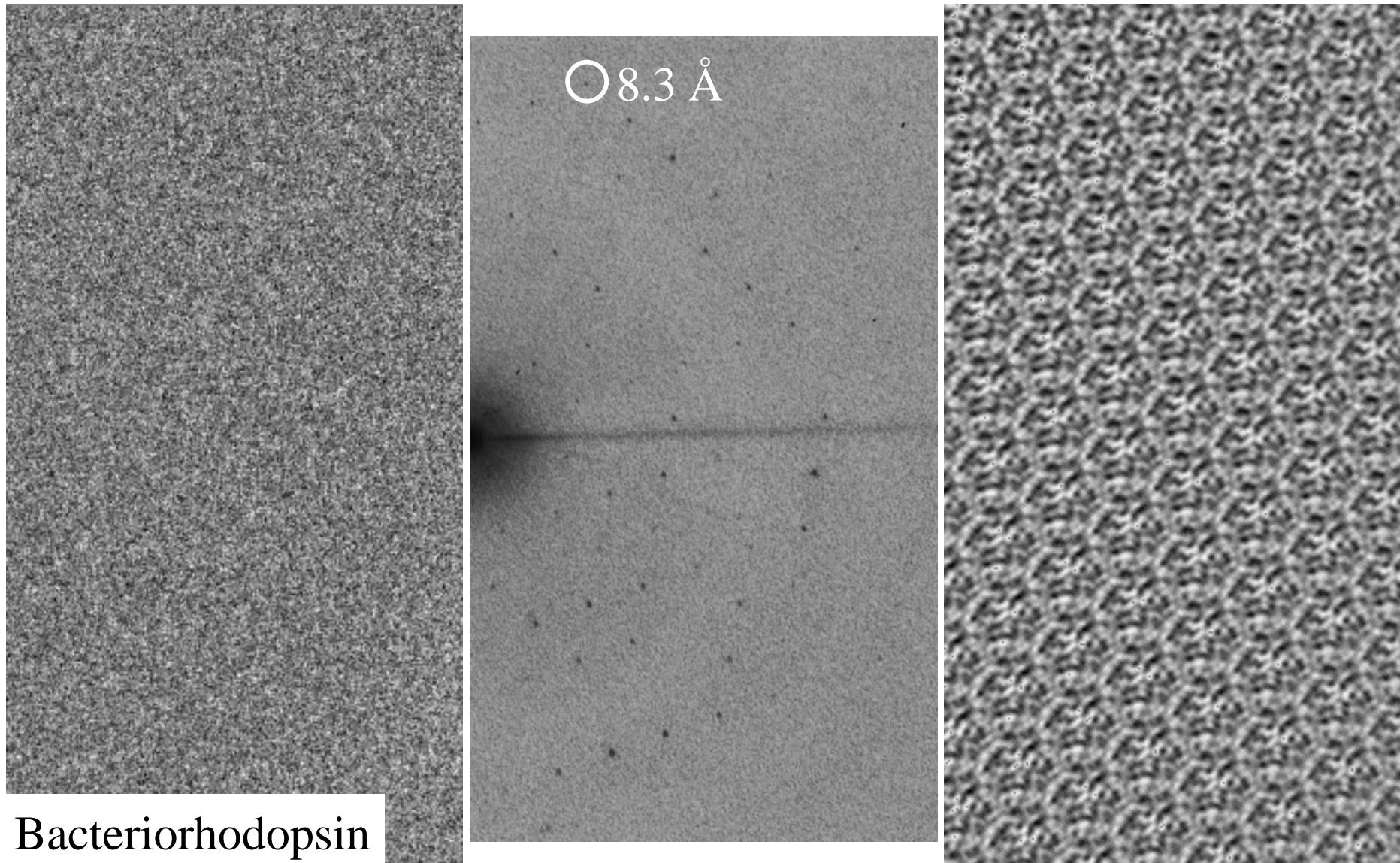
King Richard hath decreed... (QRB, 1995)

- Use 5 e⁻ per Å²
 - Demand a signal-to-noise ratio of 9 or better
 - Aim for 3 Å resolution
- ② Thou shall need to image 13,000 molecules
② For 6 Å, thou shall need only 7,000 images

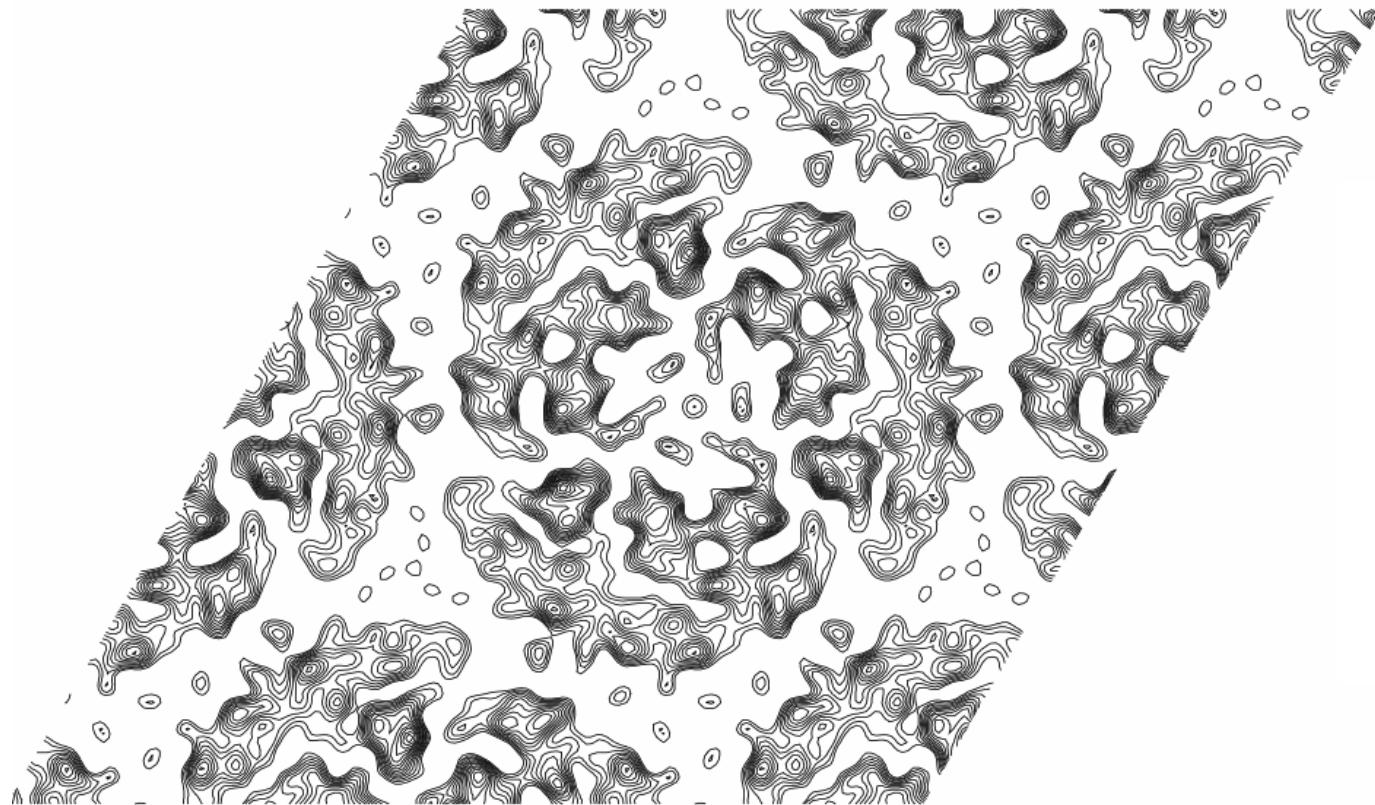
Resolving Power



Protein Crystals

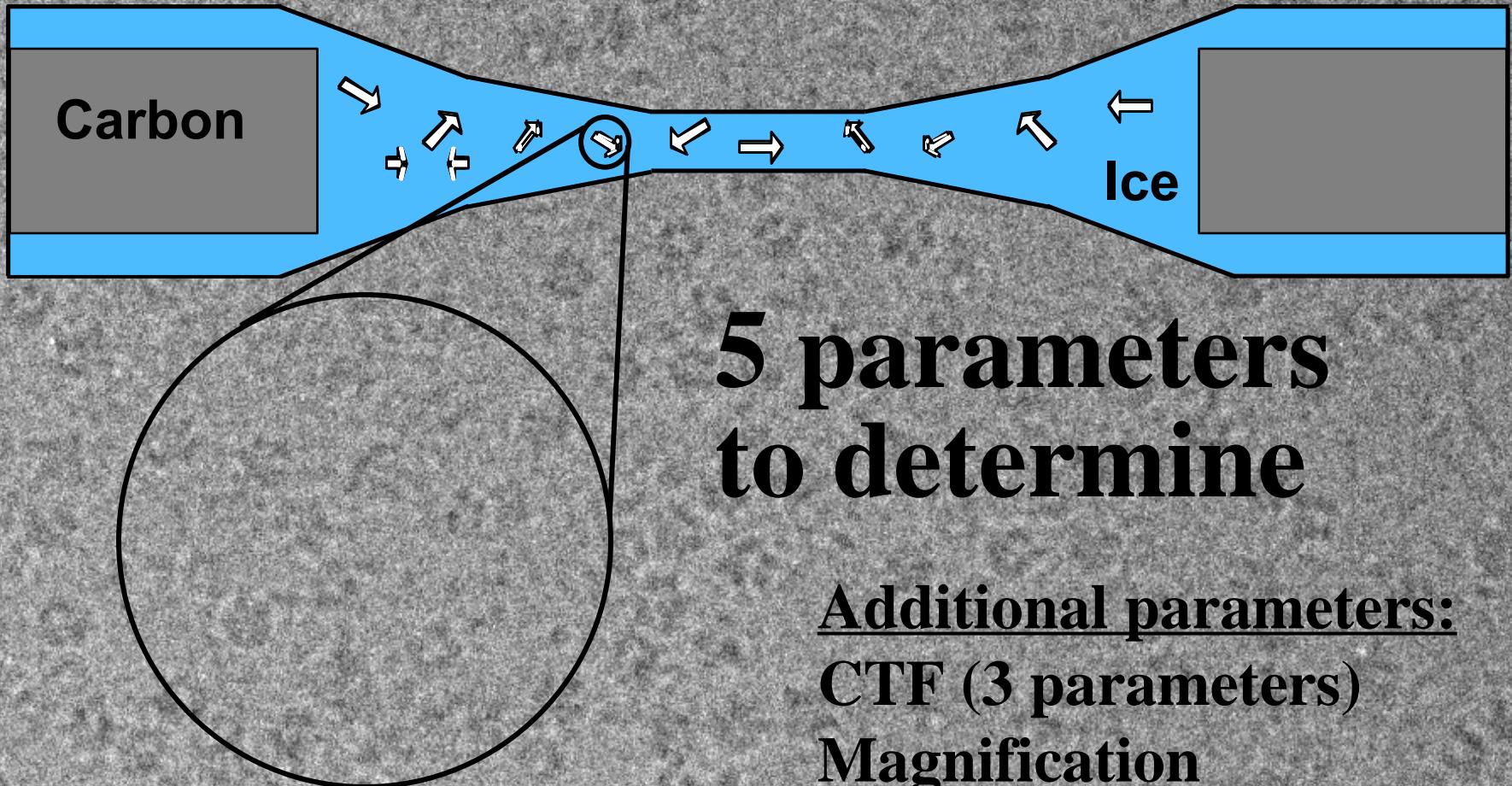


Purple Membrane



2.6 Å resolution

The Puzzle



5 parameters
to determine

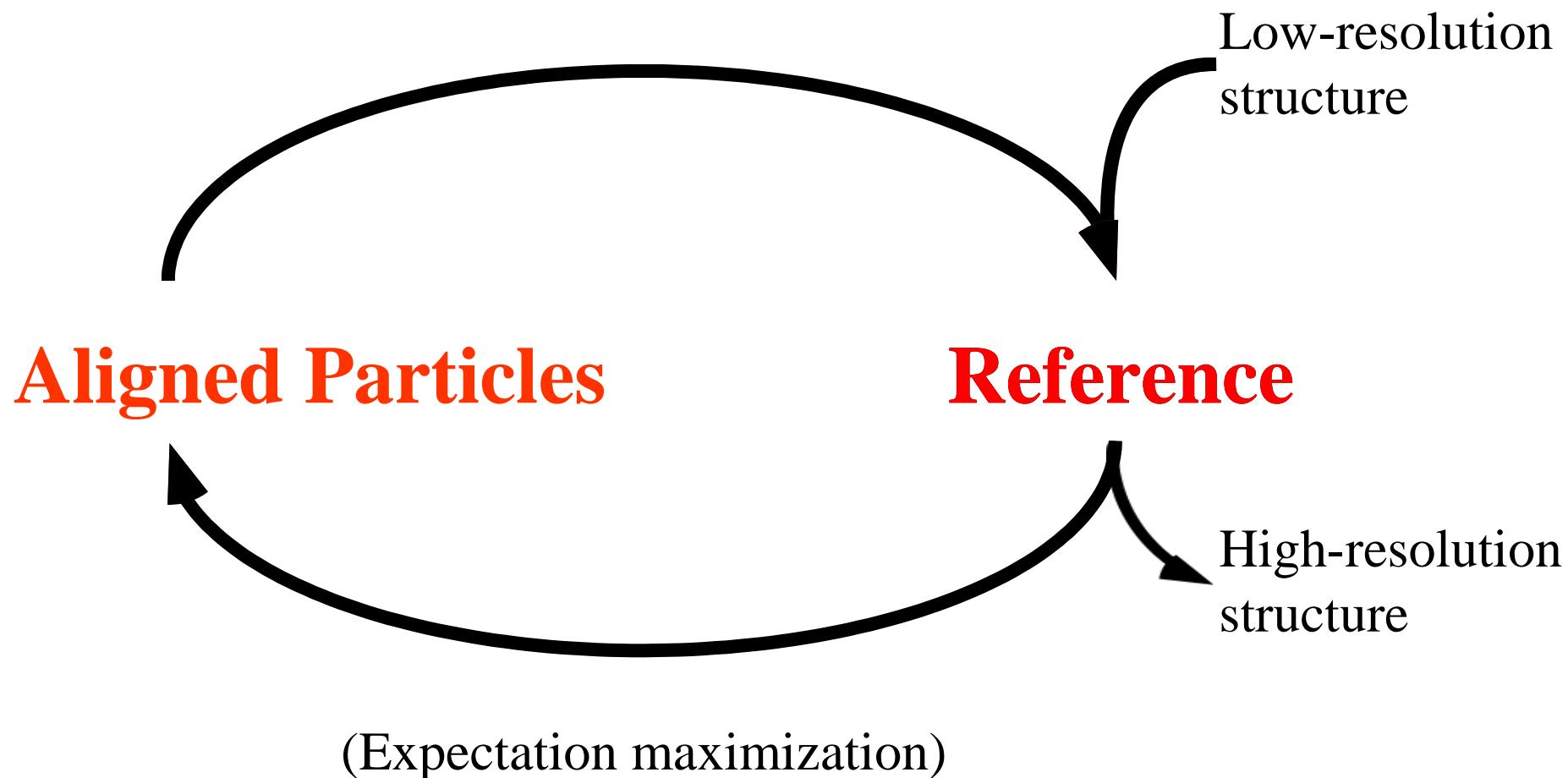
Additional parameters:
CTF (3 parameters)
Magnification
Beam Tilt (2 parameters)

200Å

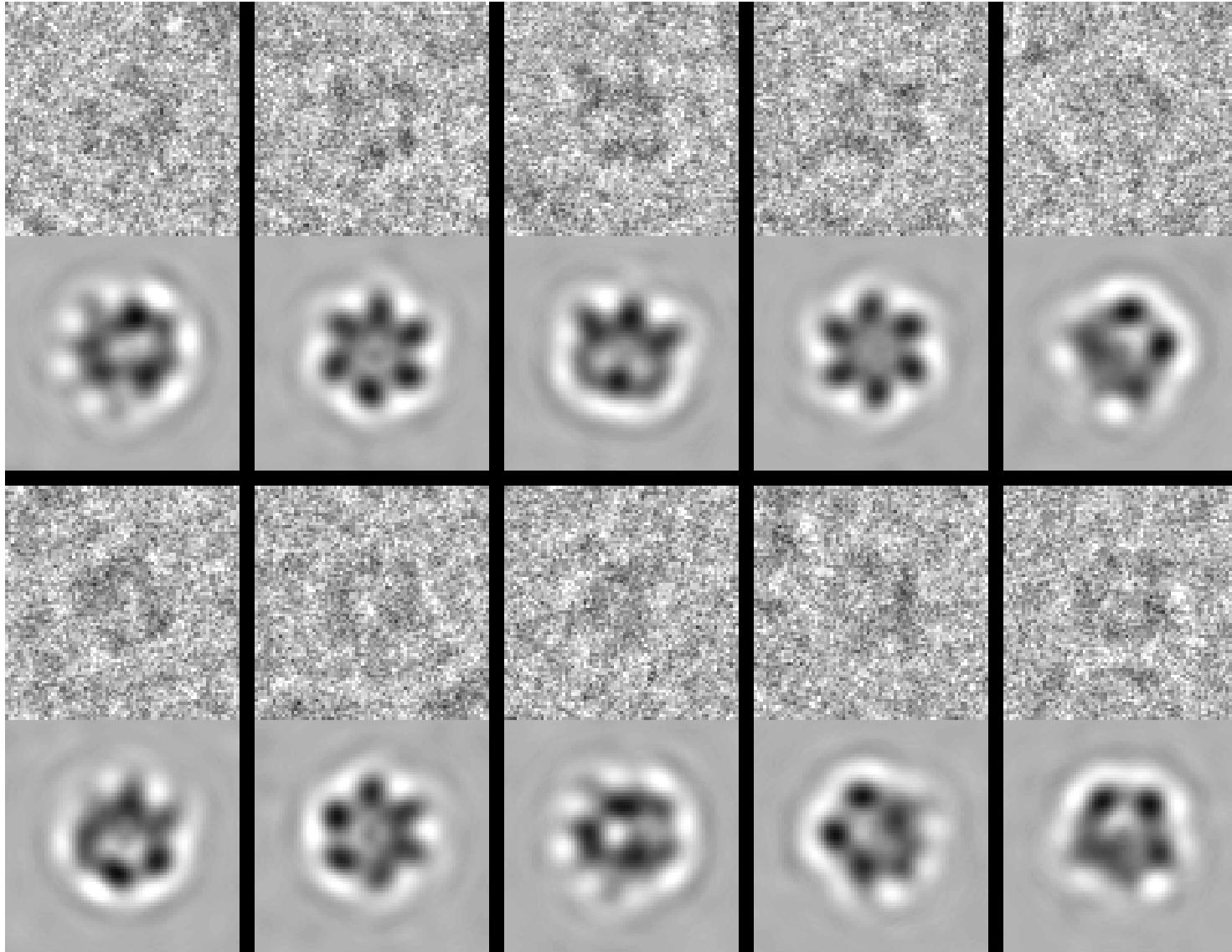
A Crazy Idea

- Assume reliable resolution measure
- Search entire parameter space for highest resolution
- Given enough images, atomic resolution is reached
- Example:
 - 3 angles, 1 deg step; two coordinates, 1 pixel step:
 $360 \times 360 \times 360 \times 100 \times 100 = 5 \times 10^{11}$
 - 13000 particles: $(5 \times 10^{11})^{13000}$ structures to search
- This is a big number!

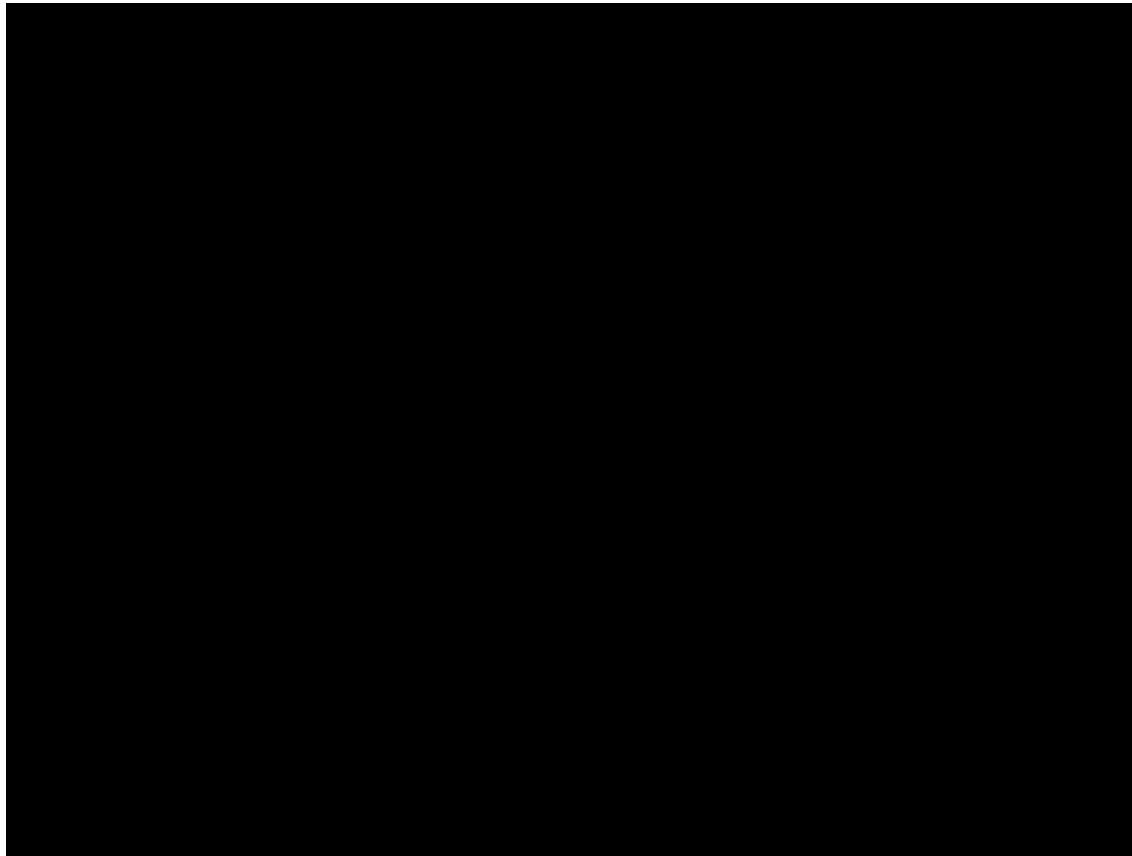
Refinement



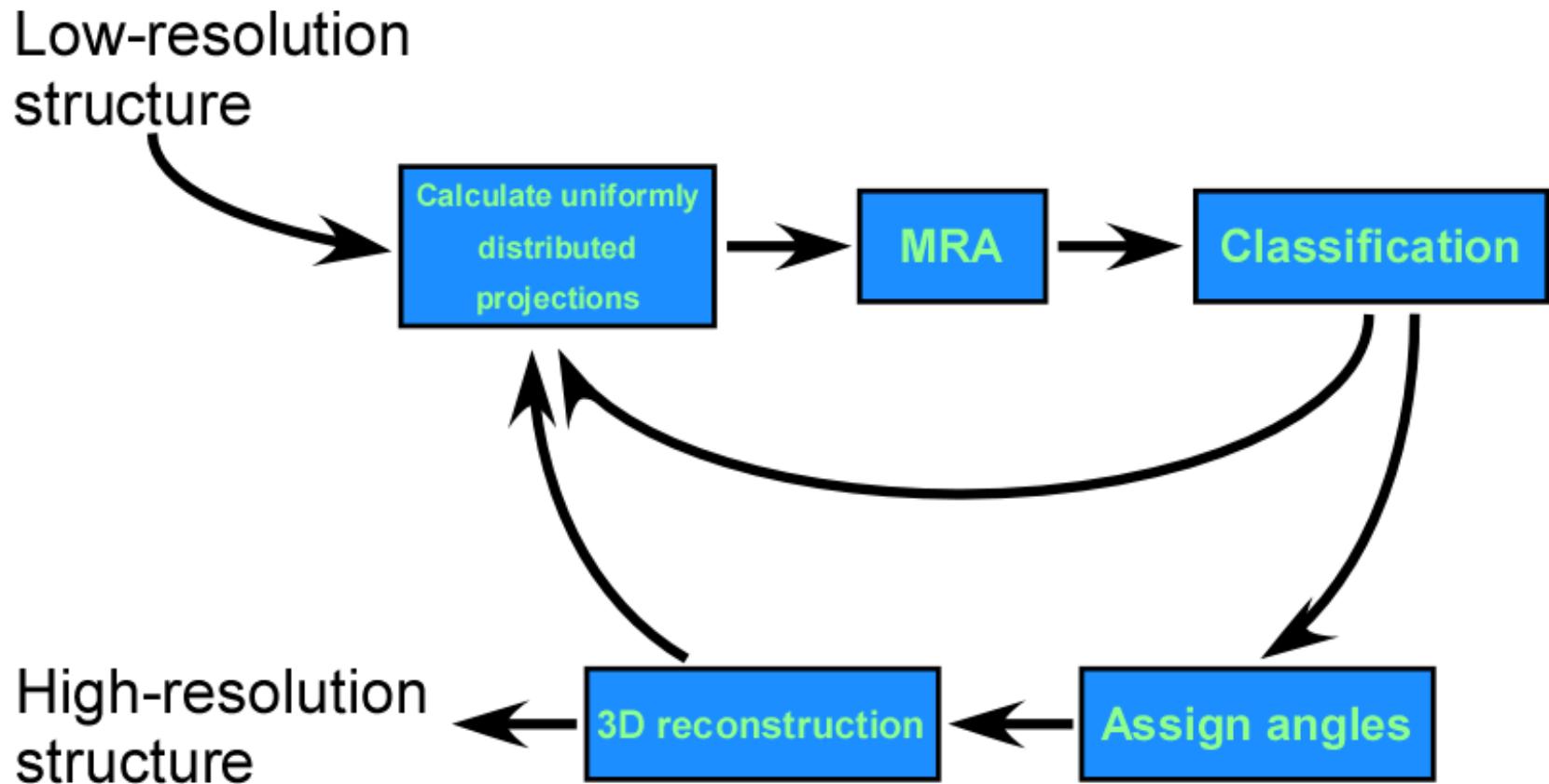
Strategy 1: Projection Matching



Strategy 2: Alignment in Reciprocal Space



Strategy 3: MRA and Classification



Strategy 4: Maximum Likelihood

**Structure for
 $n+1$ iteration**

$$A^{(n+1)} = \frac{1}{N} \sum_{i=1}^N \frac{\int X_i(\phi) p_i(\phi, \Theta^{(n)}) d\phi}{\int p_i(\phi, \Theta^{(n)}) d\phi}$$

**Probability
function**

$$p_i(\phi, \Theta) = \left(\frac{1}{\sqrt{2\pi}\sigma} \right)^M \exp \left[-\frac{|X_i(\phi) - A|^2}{2\sigma^2} \right] f(\phi | \Theta)$$

X_i : i th image

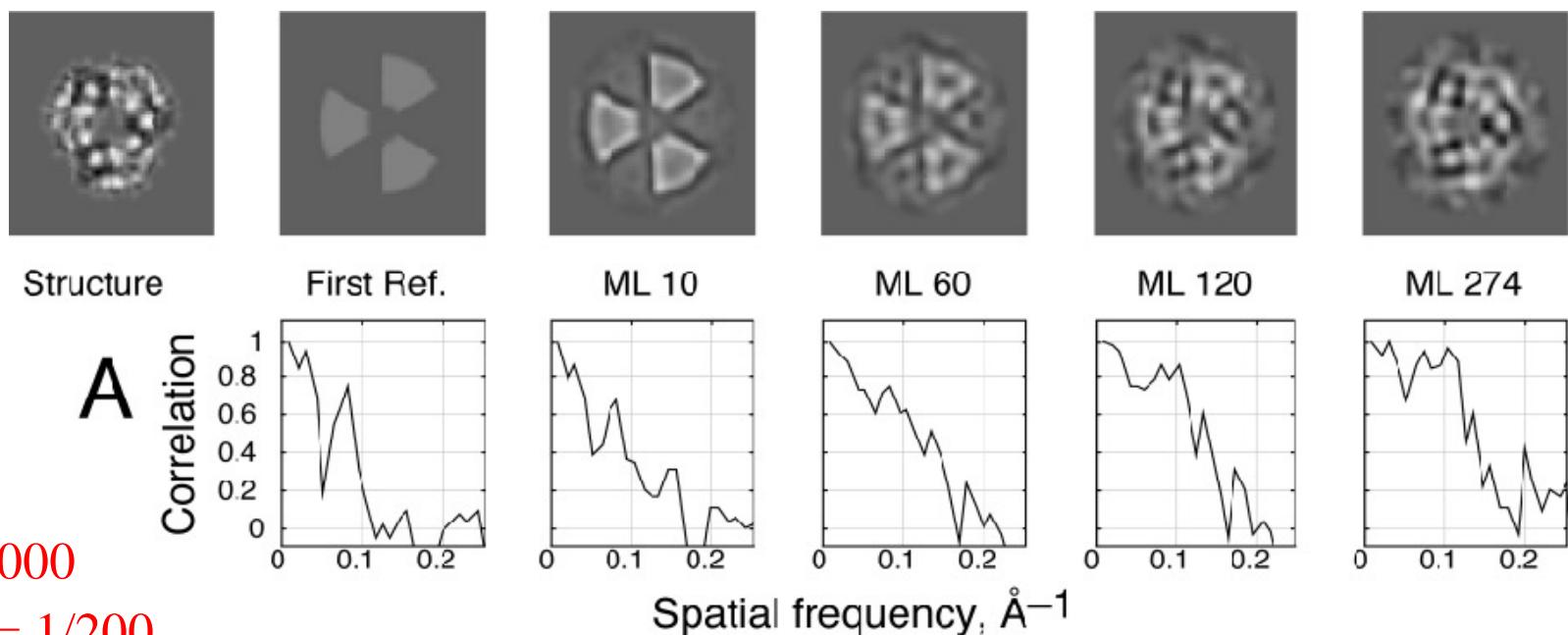
N : # of images

ϕ : alignment parameters

Θ : model parameters

σ : noise in images

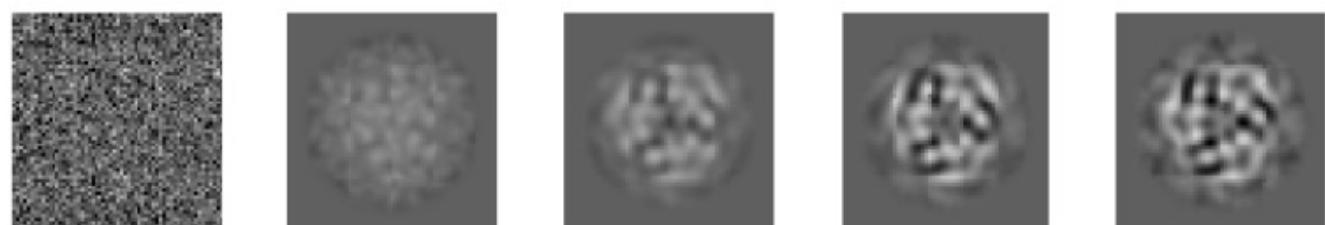
f : positional probab.



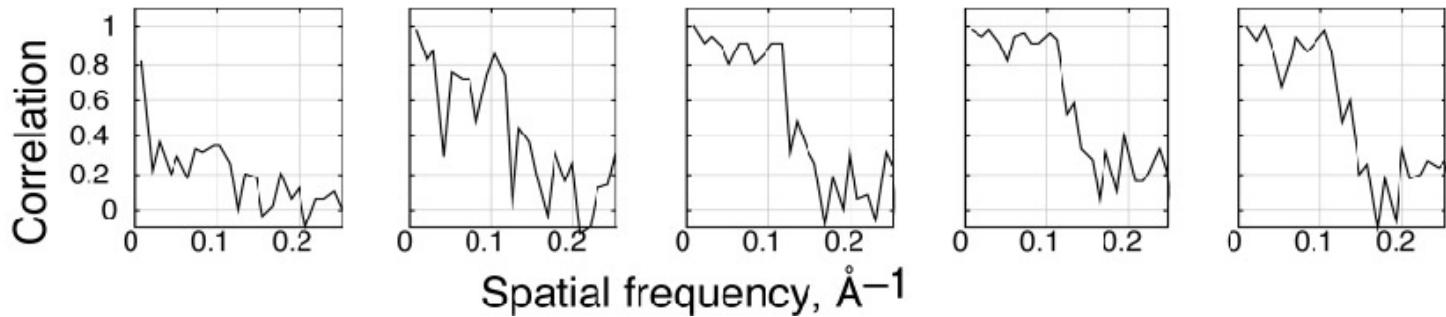
N = 4000

SNR = 1/200

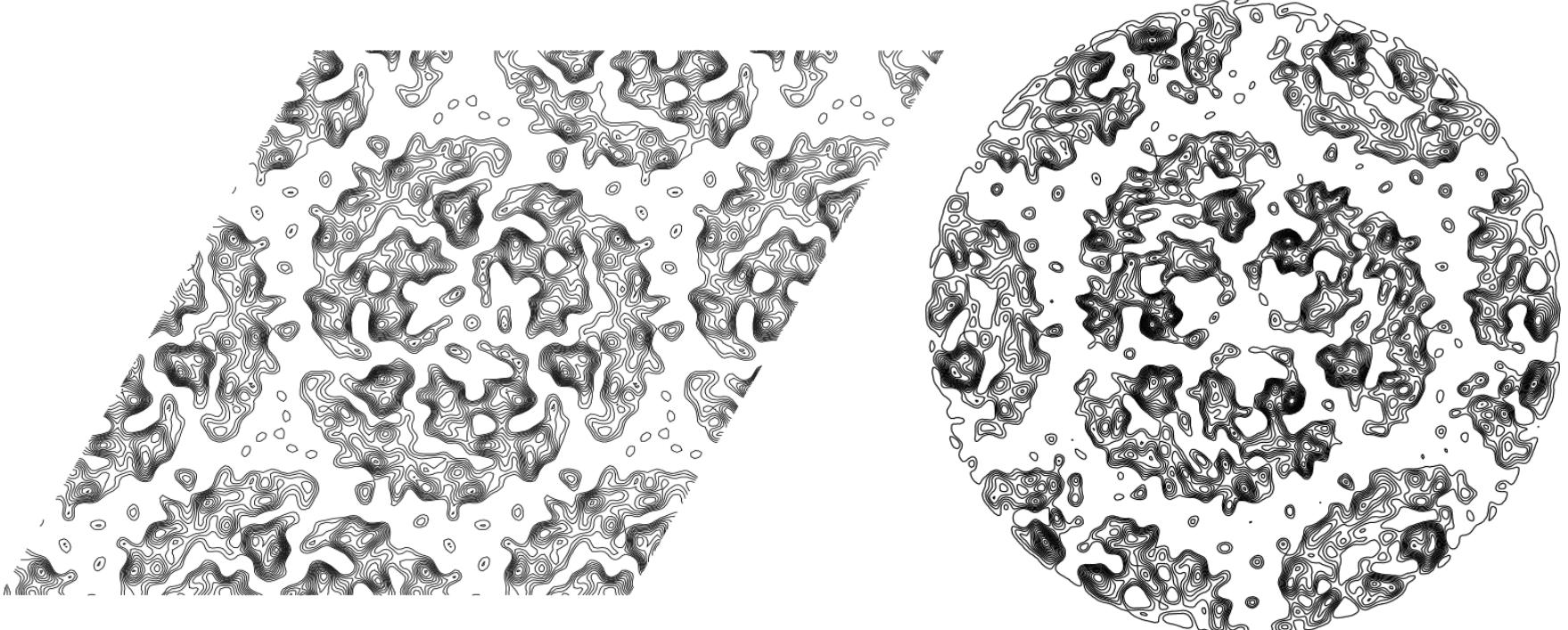
Maximum likelihood alignment



B



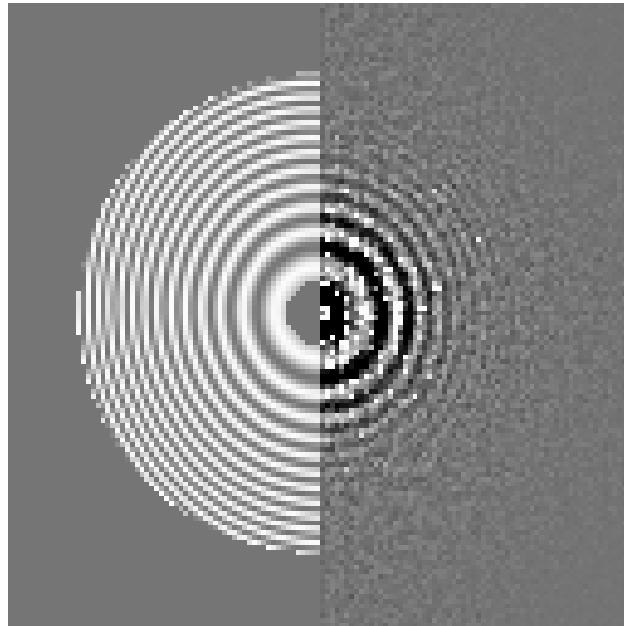
ML processing of 2D crystals



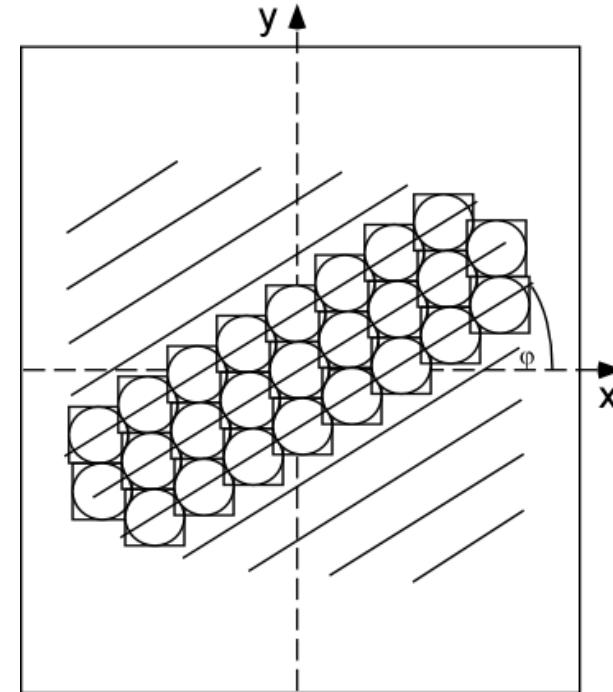
Crystallography

Alignment of
individual unit cells
using ML approach

Defocus/Astigmatism and Magnification

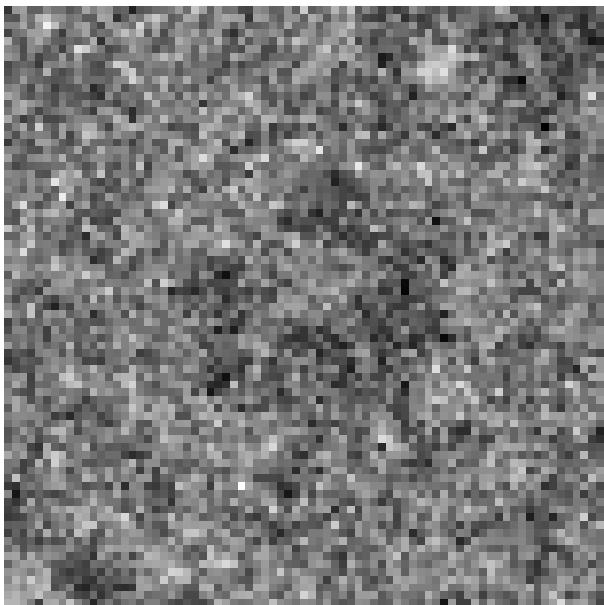


CTFFIND3

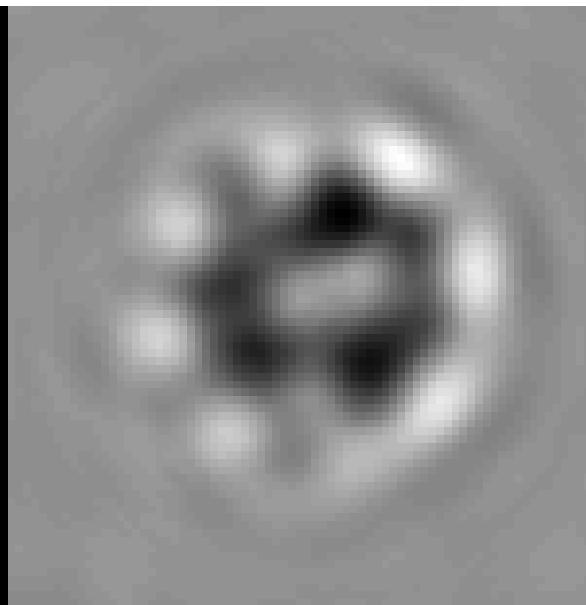


CTFTILT

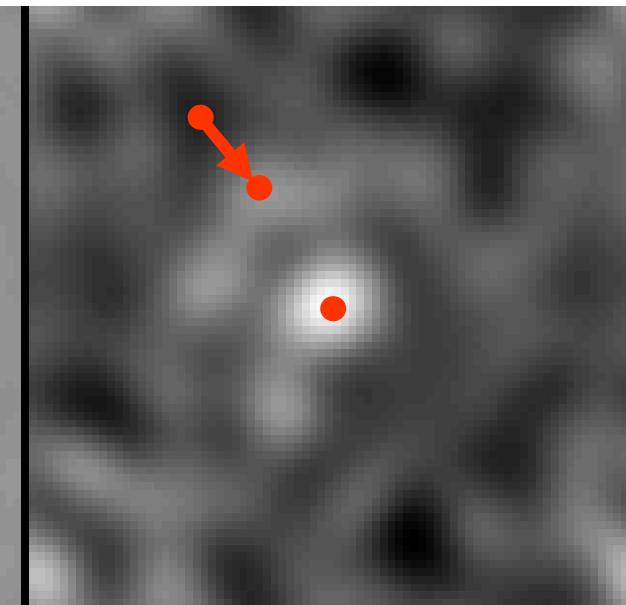
Problem 1: Local Optima



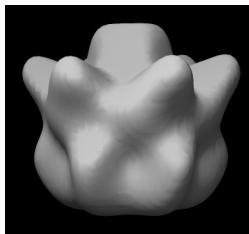
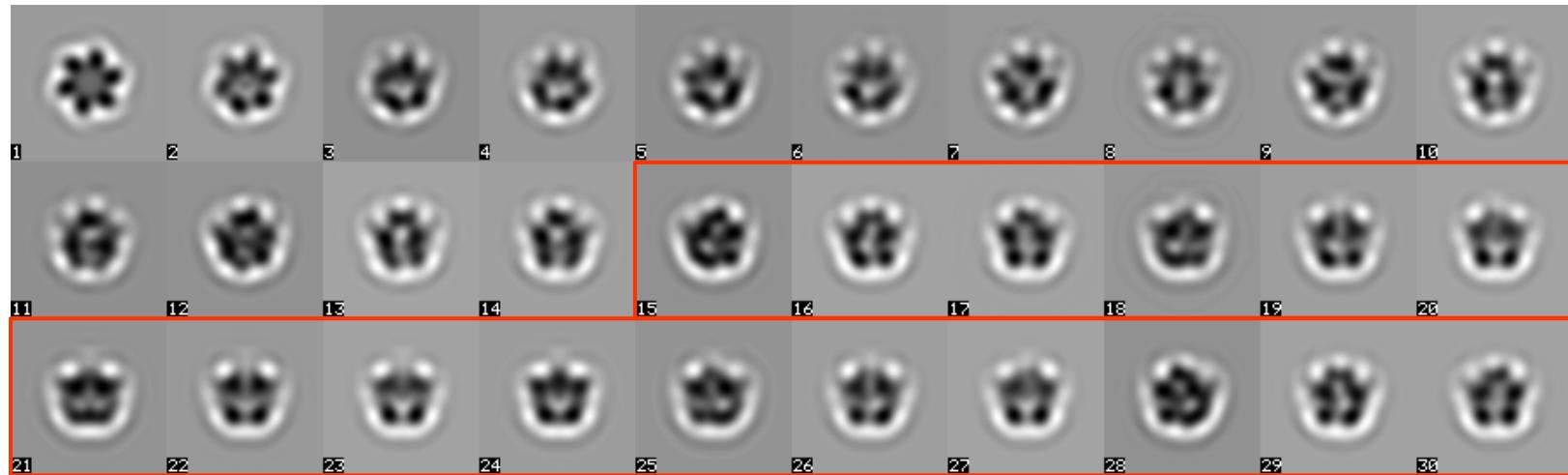
Particle



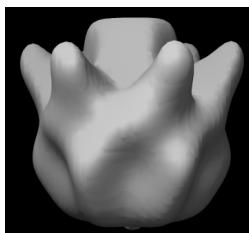
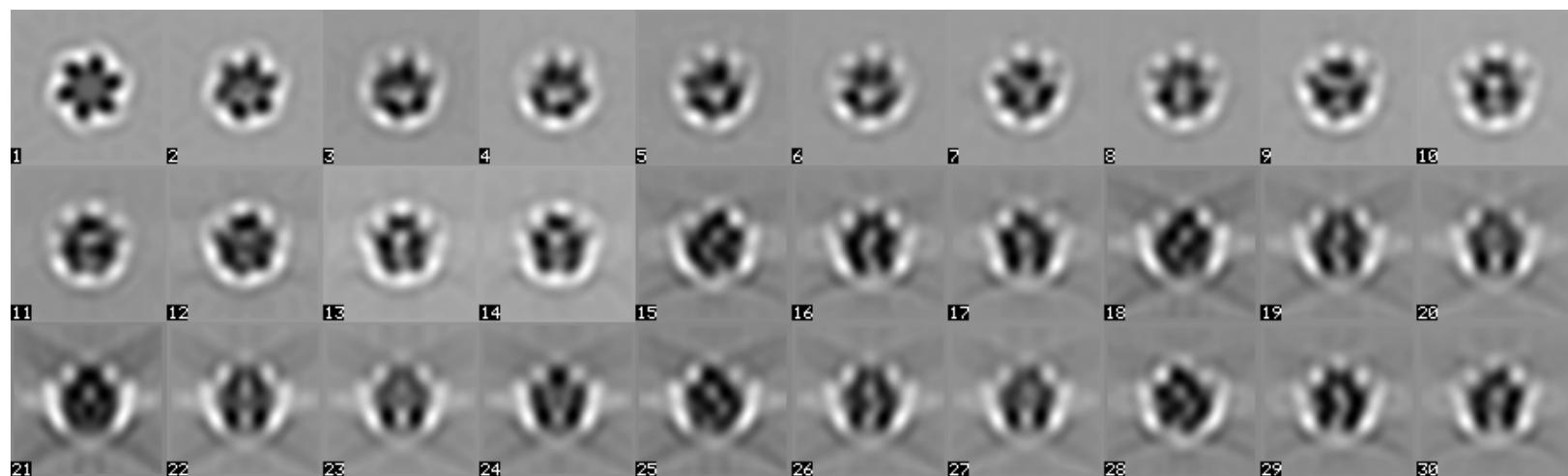
Reference



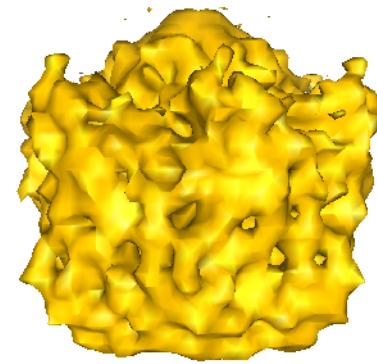
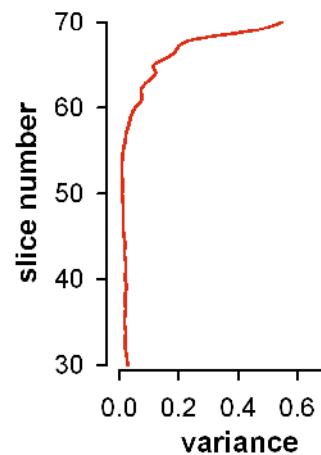
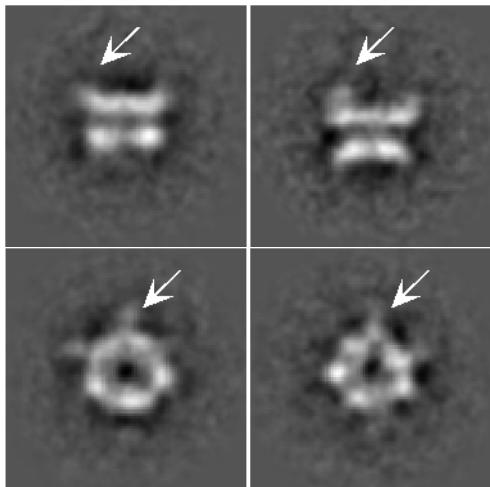
Problem 2: Missing Views



> 60°



Problem 3: Heterogeneity



- Misalignment of particles
- Lower resolution in disordered regions
- Loss of features

Classification Using ML

**Structure for
 $n+1$ iteration**

$$A_k^{(n+1)} = \frac{1}{\sum_i q_i^k(\Theta)} \sum_{i=1}^N \frac{\int X_i(\phi) p_i^k(\phi, \Theta^{(n)}) d\phi}{\sum_k \int p_i^k(\phi, \Theta^{(n)}) d\phi}$$

**Probability
function**

$$p_i^k(\phi, \Theta) = \left(\frac{1}{\sqrt{2\pi}\sigma} \right)^M \exp \left[-\frac{|X_i(\phi) - A_k|^2}{2\sigma^2} \right] f(\phi | \Theta)$$

**Probability
for class k**

$$q_i^k(\Theta) = \int p_i^k(\phi, \Theta) d\phi$$

X_i : i th image

N : # of images

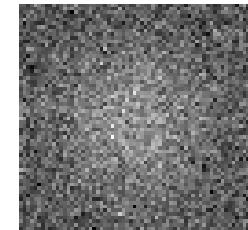
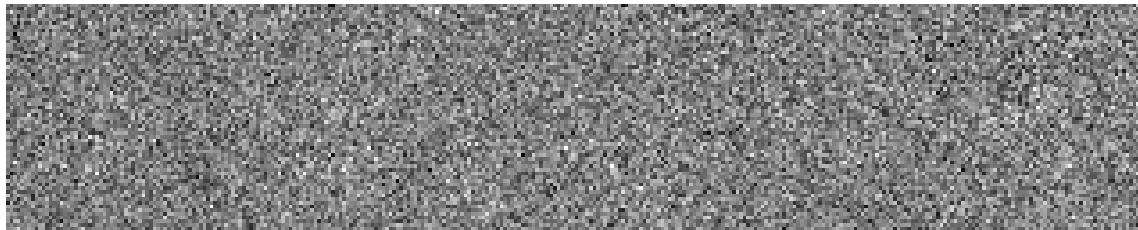
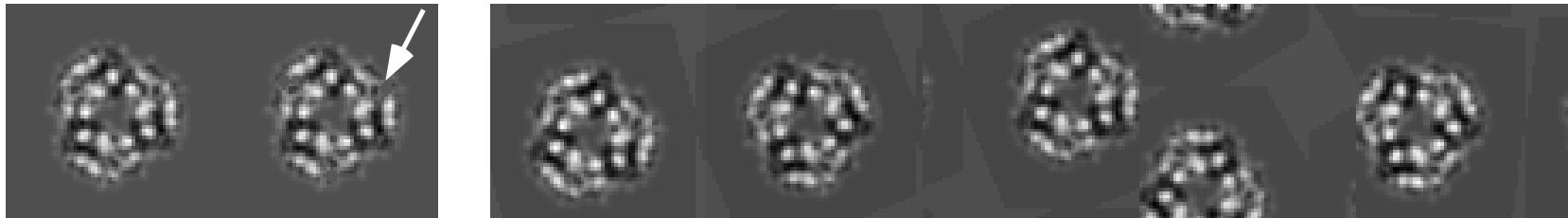
ϕ : alignment parameters

Θ : model parameters

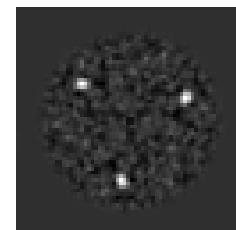
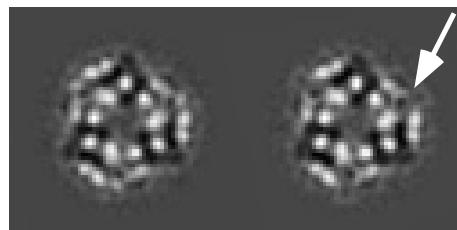
σ : noise in images

f : positional probab.

Classification Using ML

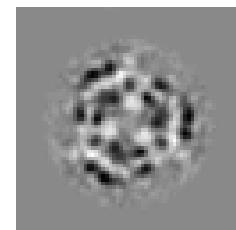
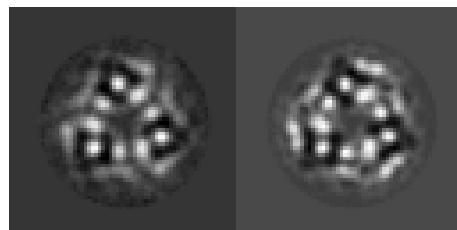


$\text{SNR} = 1/50$
 $N = 2000$

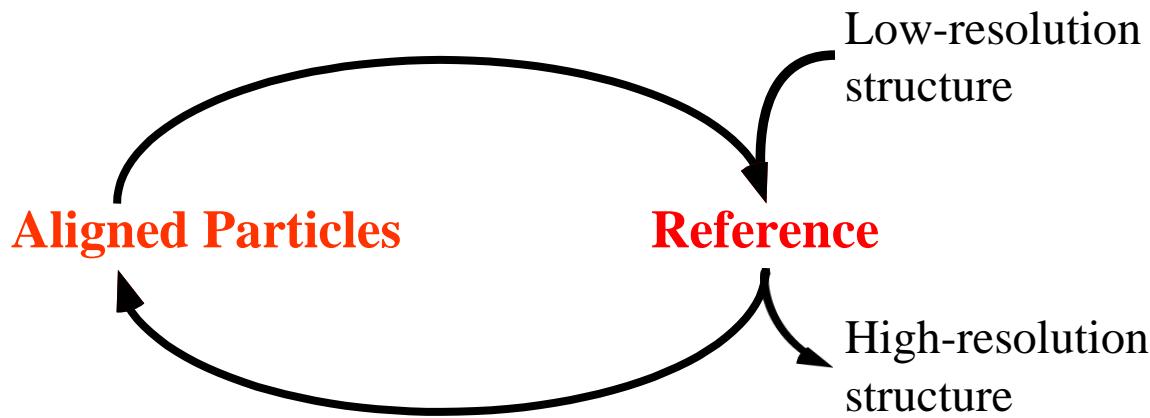


Difference
map

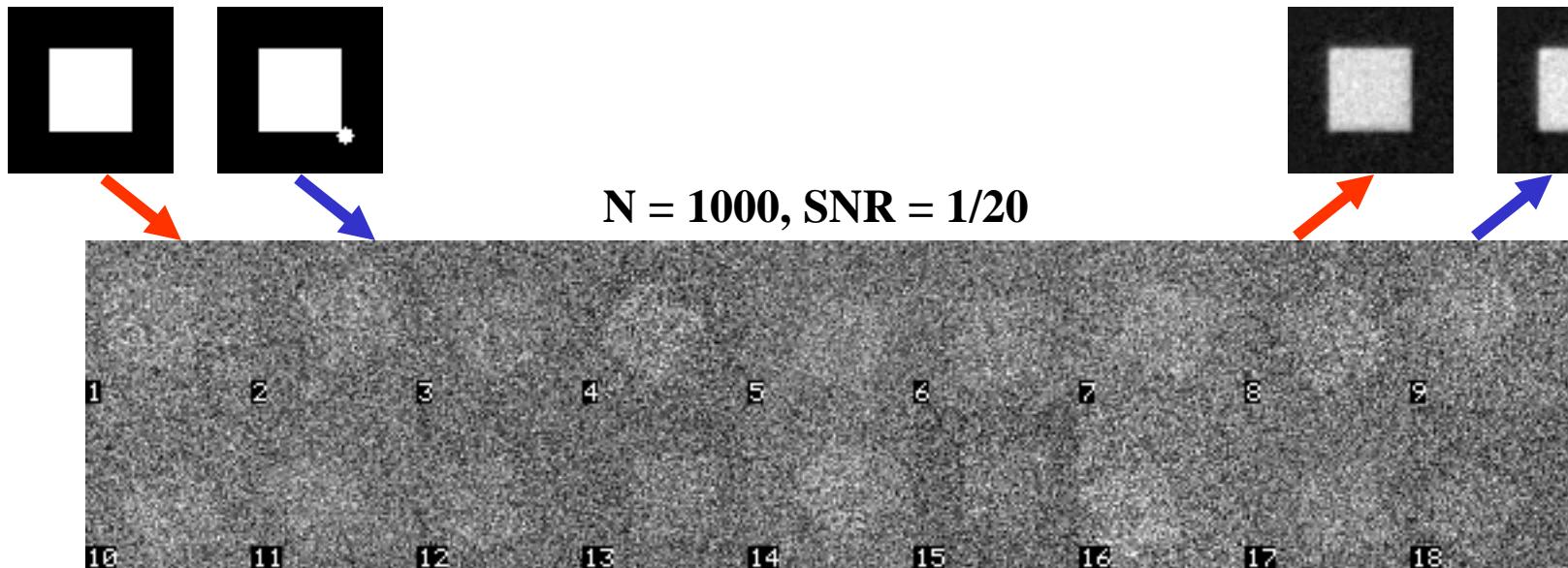
Correlation
alignment



Problem 4: Processing Artifacts



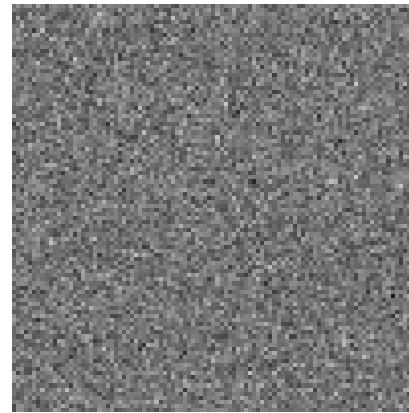
- Interpolation errors
- Masking
- Negative B-factor
- ...



Problem 5: Noise Bias



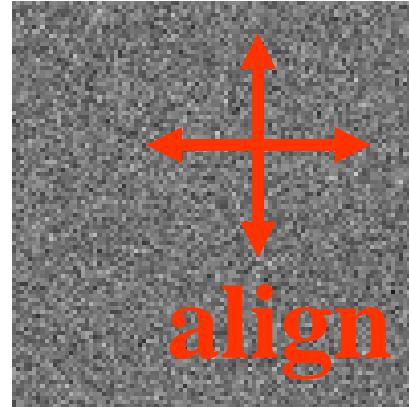
\otimes



= 0 on average



\otimes



> 0

for 64x64 image:
average
correlation = 0.064

Seeing is NOT Always Believing



100 Images

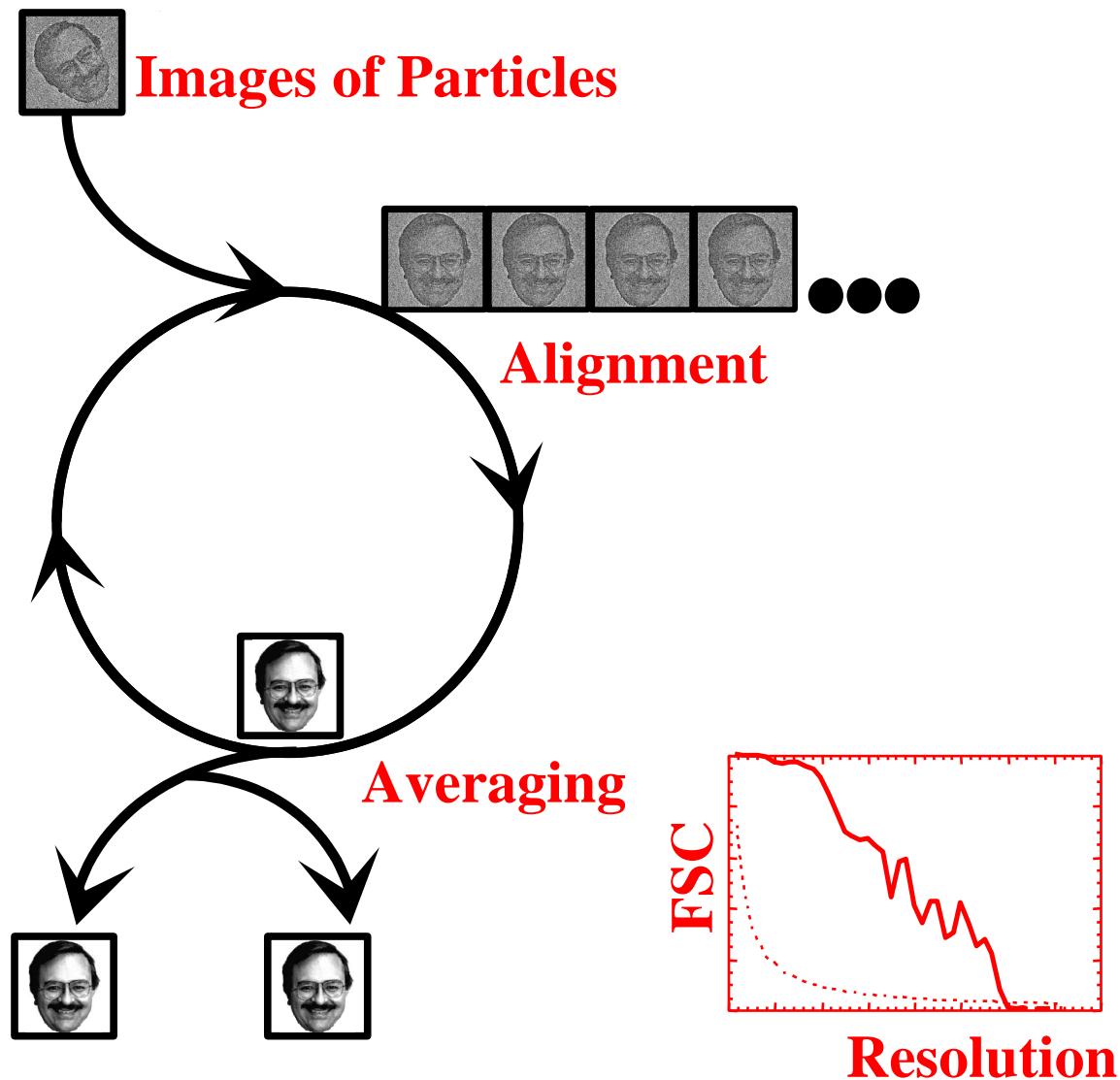


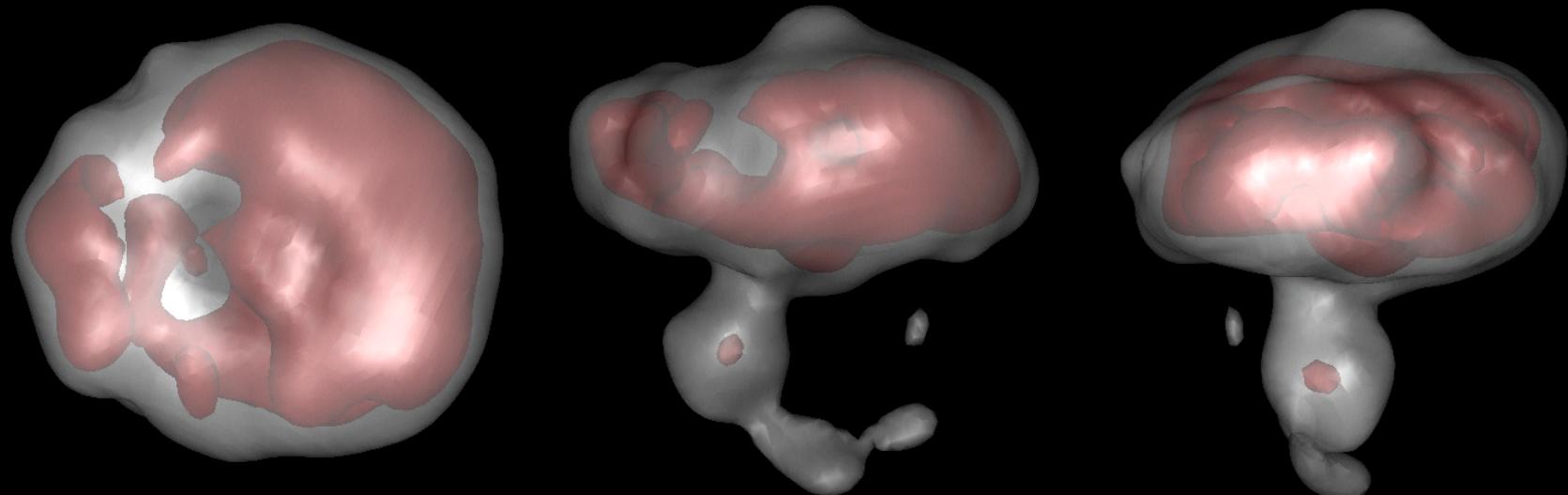
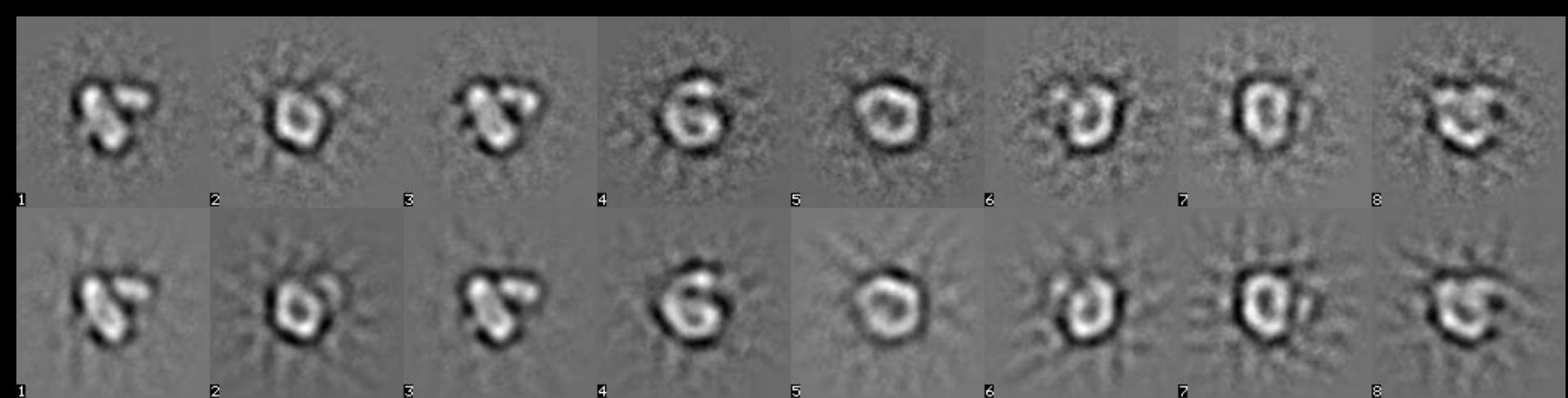
1000 Images



Reference

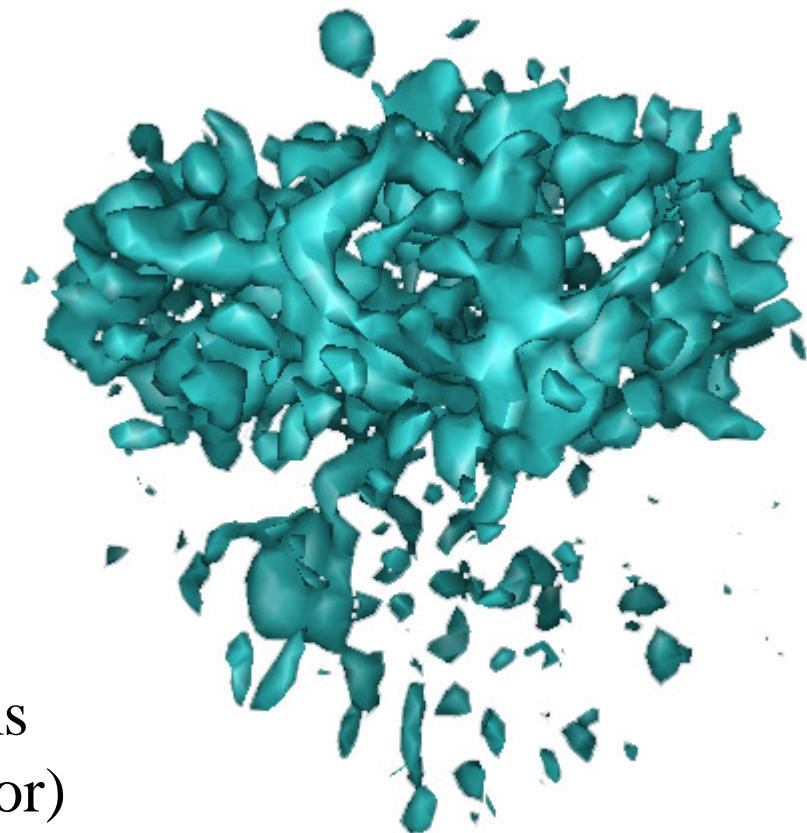
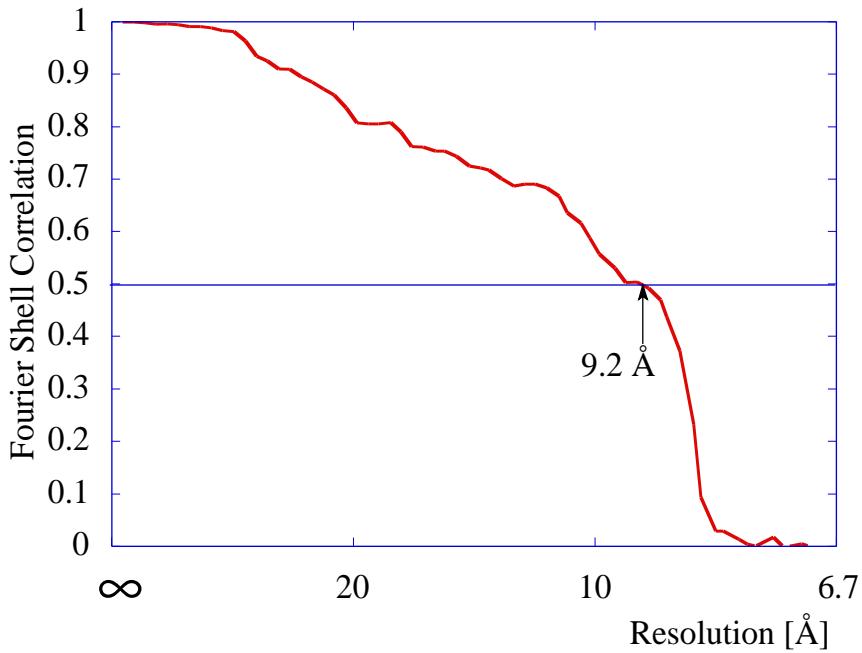
Resolution Measurement





100 Å

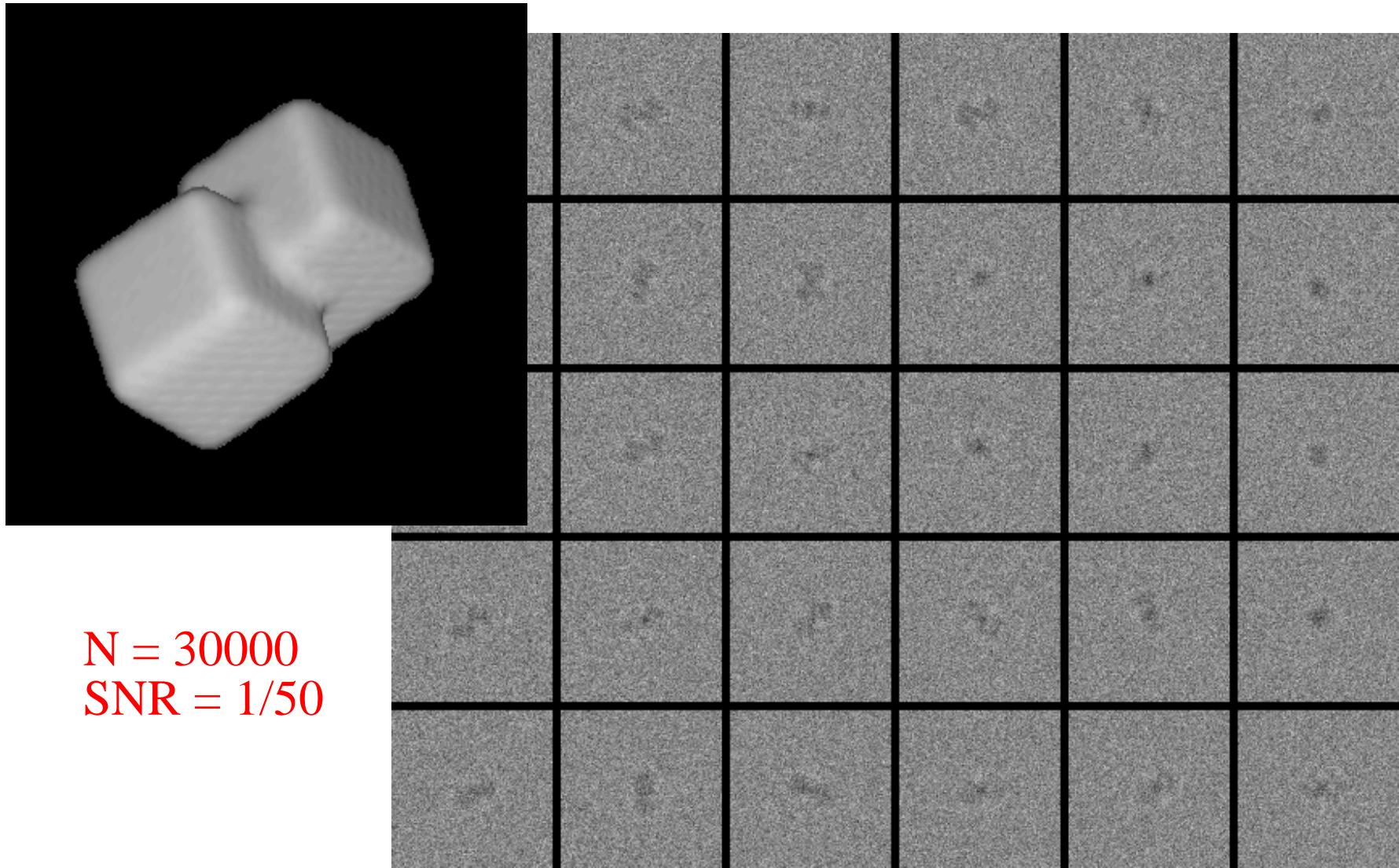
Swiss Cheese



Dangerous:

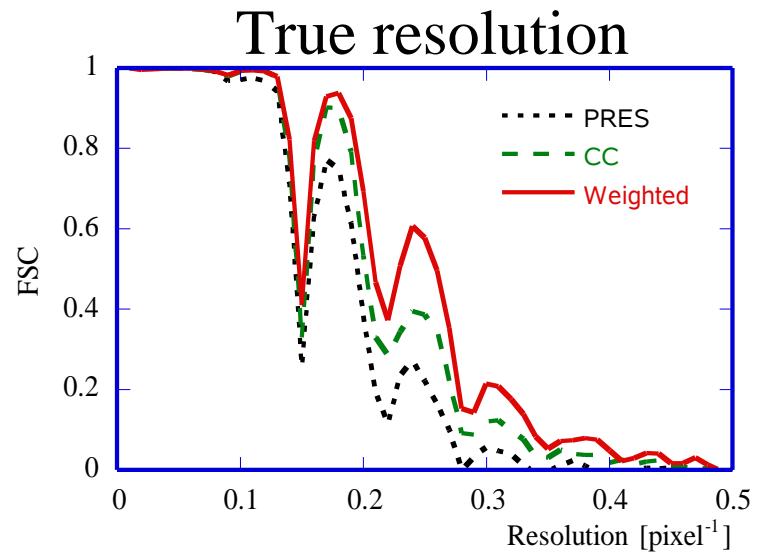
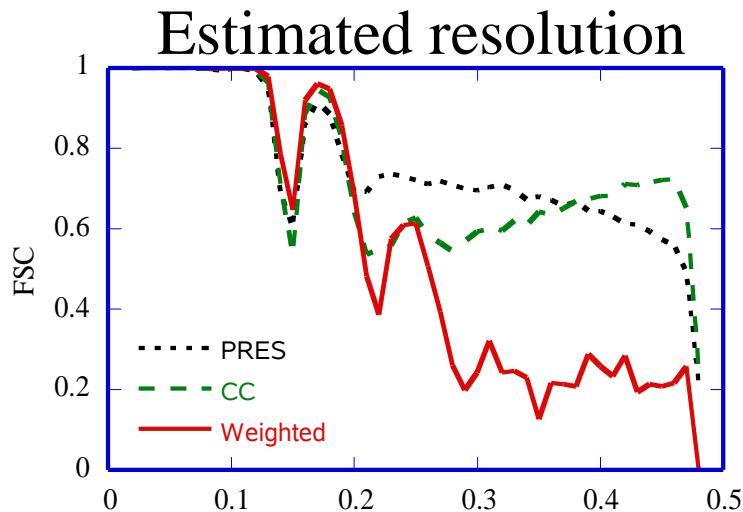
Boosting of high-resolution terms
(application of a negative B-factor)

Gedanken Experiments



$N = 30000$
 $\text{SNR} = 1/50$

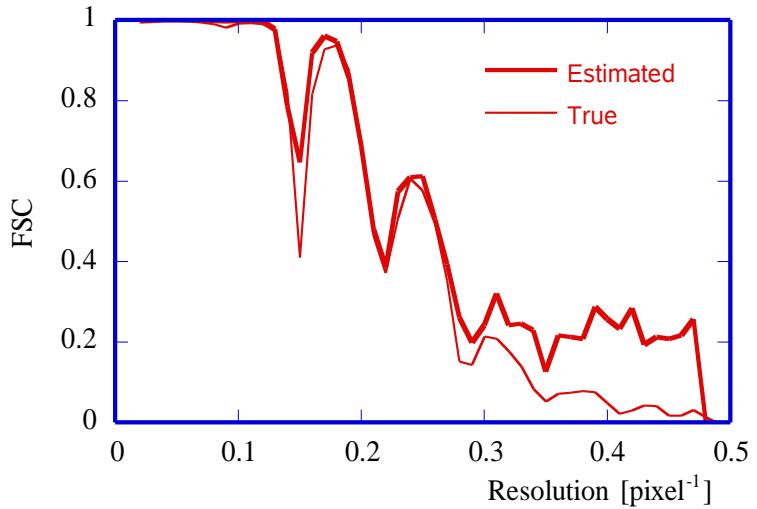
Weighted Correlation



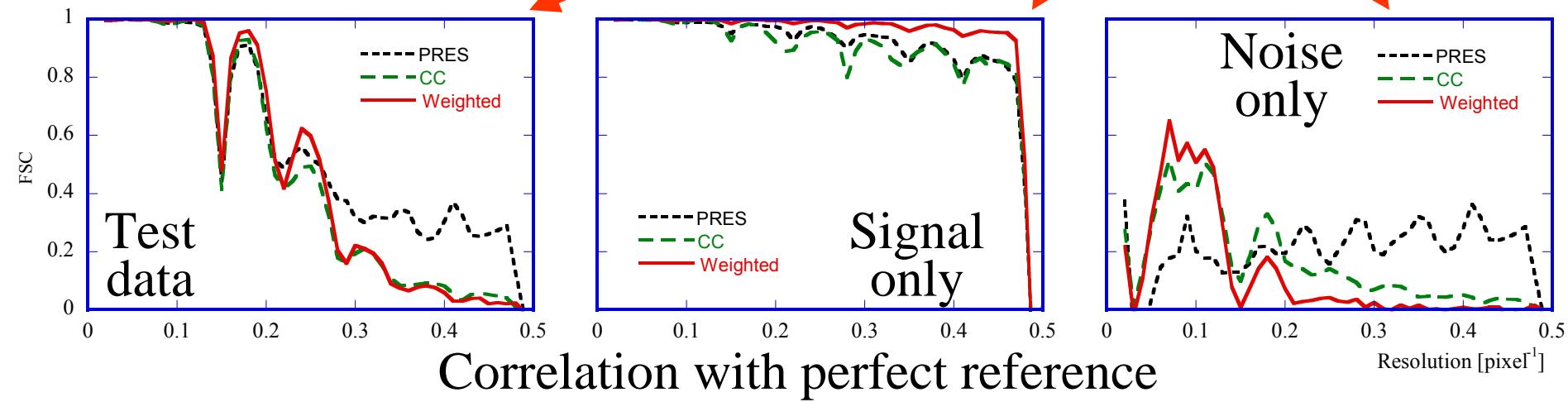
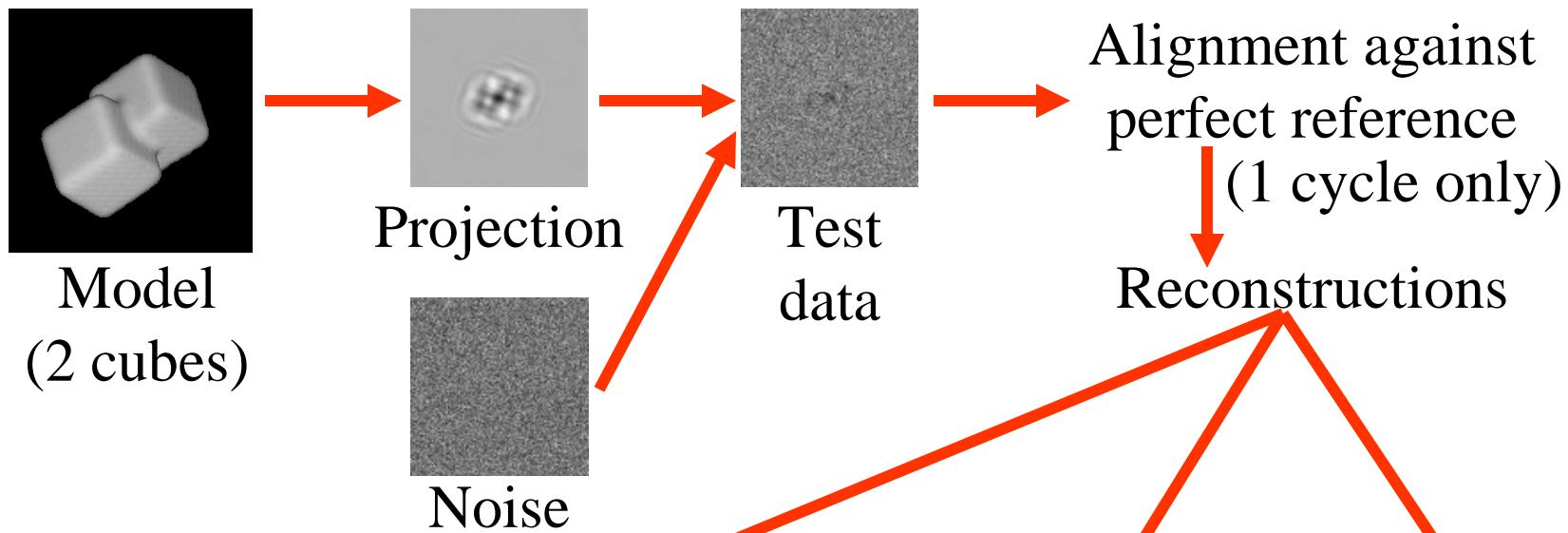
$$\text{PRES}(X, Y) = \frac{\sum_{k \in [0, 0.5]} \Delta\Phi_{X,Y}(\mathbf{k}) |F_X(\mathbf{k})|}{\sum_{k \in [0, 0.5]} |F_X(\mathbf{k})|}$$

$$\text{CC}(X, Y) = \frac{\sum_{k \in [0, 0.5]} F_X(\mathbf{k}) F_Y^\dagger(\mathbf{k})}{\sqrt{\sum_{k \in [0, 0.5]} |F_X(\mathbf{k})|^2 \sum_{k \in [0, 0.5]} |F_Y(\mathbf{k})|^2}}$$

$$\text{CC}_W(X, Y) \in_W (\sum_{k \in [0, 0.5]} W)_X \nparallel_k \sum_i F_X(Q_i^3) F_Y^\dagger(\mathbf{k})$$

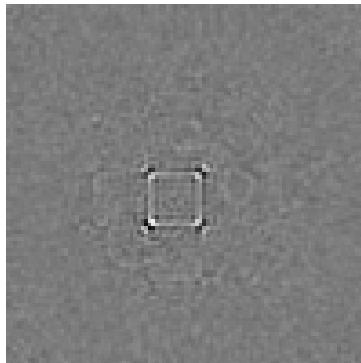


Noise Bias

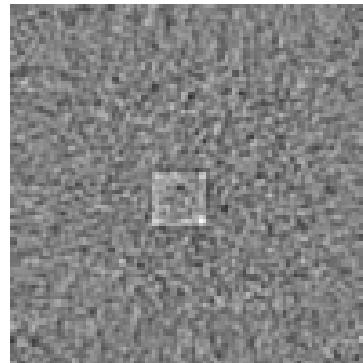


Noise Reconstruction

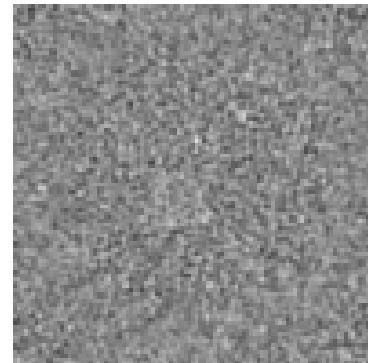
Phase
residual



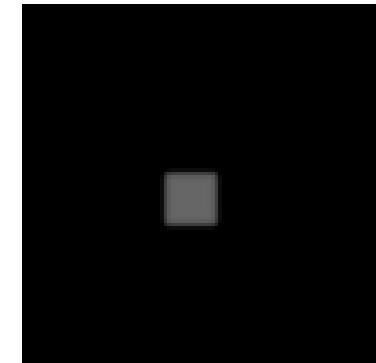
Linear
correlation
coefficient



Weighted
correlation
coefficient

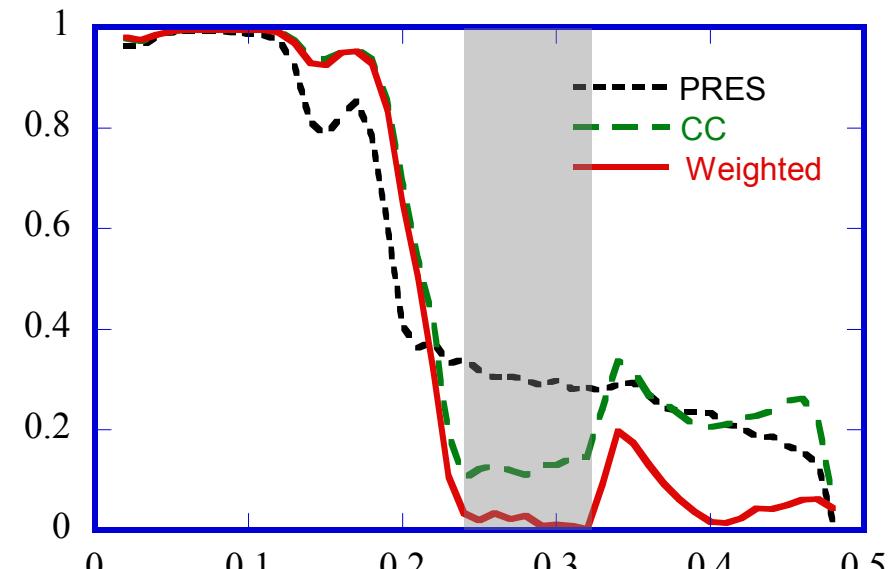
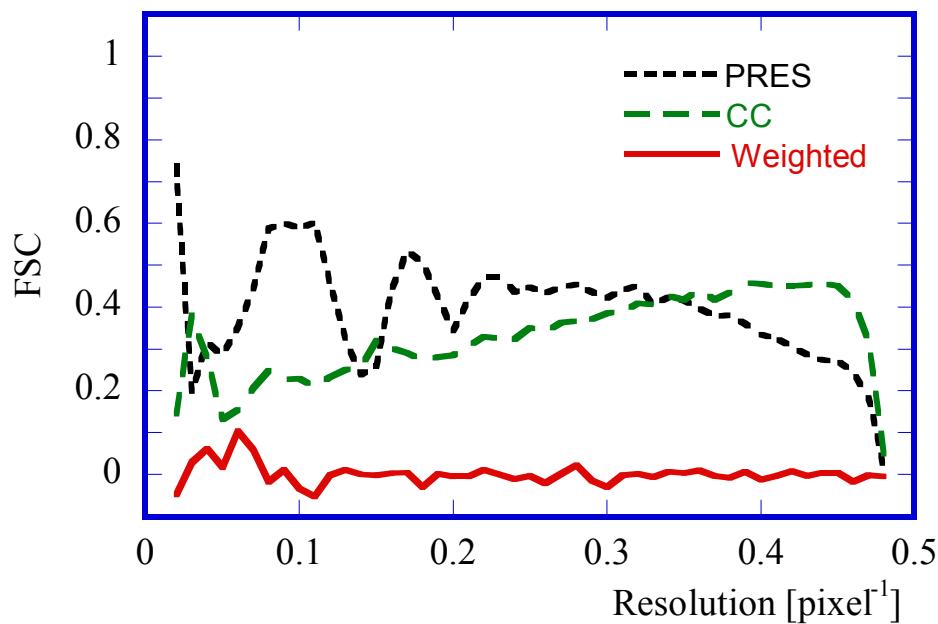


Reference



Coherence Constraint

$$\text{CC}_W(X, Y) = \sum_i |\text{CC}|_i^3$$



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- Ca Channel
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(Axel Brünger)
David DeRosier

HHMI, NIH, NSF

