

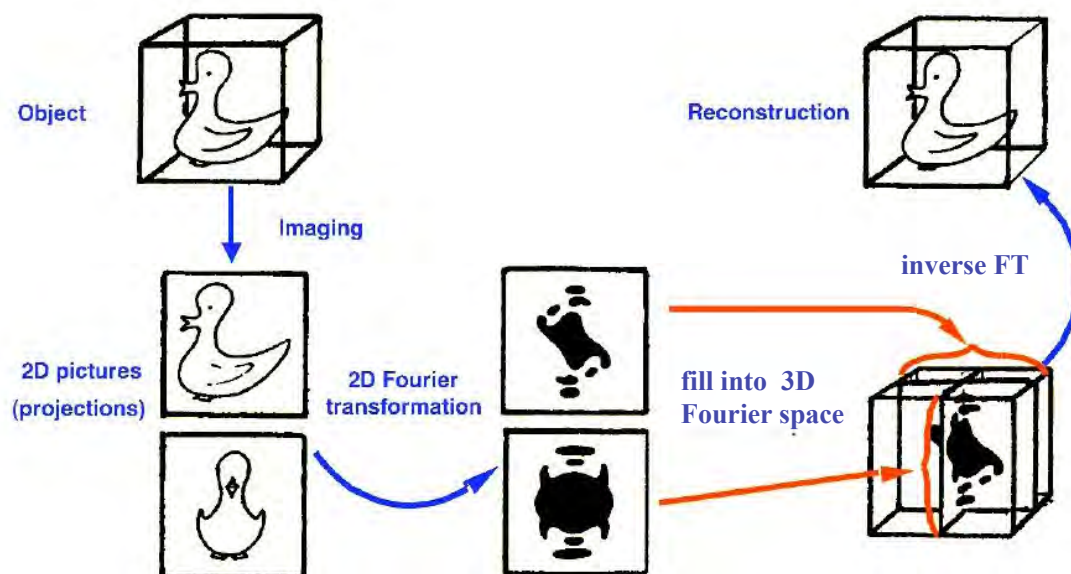
# 3D Single Particles Approaches to 2D Crystals

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**3D Reconstruction:** Assign the Euler angles to projections, and “fill in” the 3D Fourier space.



## Challenge: extremely low signal-to-noise ratio (SNR)

- Crystal images: apply Fourier filtration to **ordered** crystals

The average structure is contained in the diffraction spots (as sharp as possible) in the Fourier transform of the image.

- Single particle images: calculate the average of **aligned** particles in the same orientation

**Crystallography method:** Unbending process

**Single particle method:** Alignment process

Crystal imperfections translate into misalignments of particles

## **Concerns about Unbending:**

- Unbending is in 2D space: sample support is lack of flatness  
(tilt angle will be slightly different for each unit cell)
- Error sources:
  - Displacement accuracy: displacement obtained by cross-correlation of the reference and the image
  - Deformation process

## **Single Particle Processing:**

- Unbending can be done in 3D space: tilt angle can be different for each unit cell
- Error sources:
  - Alignment: cross-correlation, heuristic optimization, maximum likelihood
- No deformation in the unit cells

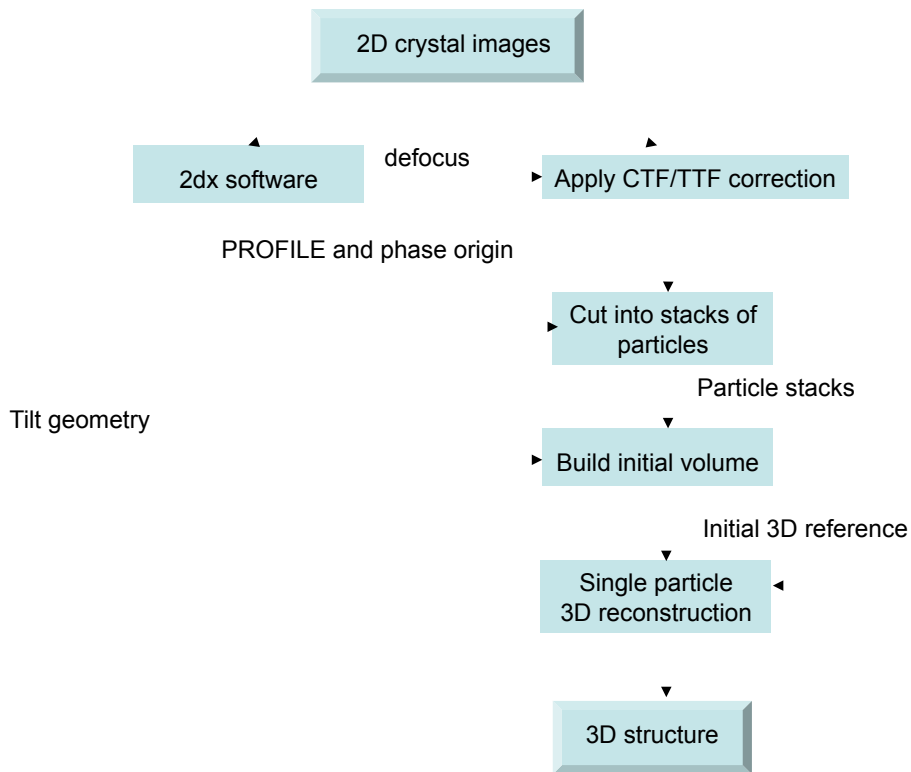
## Single Particle Approaches to Single Particles

- Classify particles by orientations
- Align and average particles within classes
- Build 3D model
- Refine 3D model

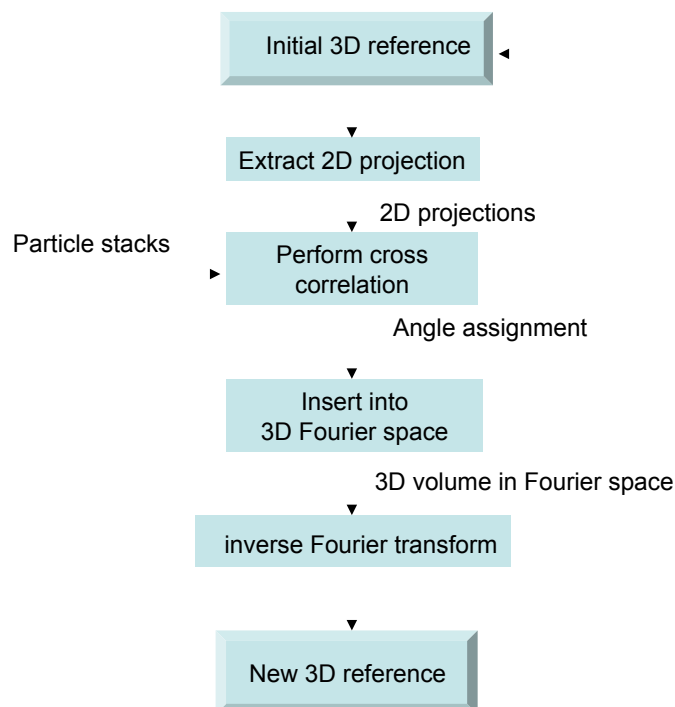
No crystal information. Not easy to get initial 3D model.

## Single Particles Approaches Applied to 2D Crystals

1. Make the best use of available information: crystal line, tilt geometry
2. Tackle the issue of lack of flatness by refining
  - Phase origin (the center of the particle)
  - Tilt geometry



## Single Particle 3D Reconstruction



**Cross-correlation alignment:**  $\phi = (\varphi, \theta, \psi, x, y)$

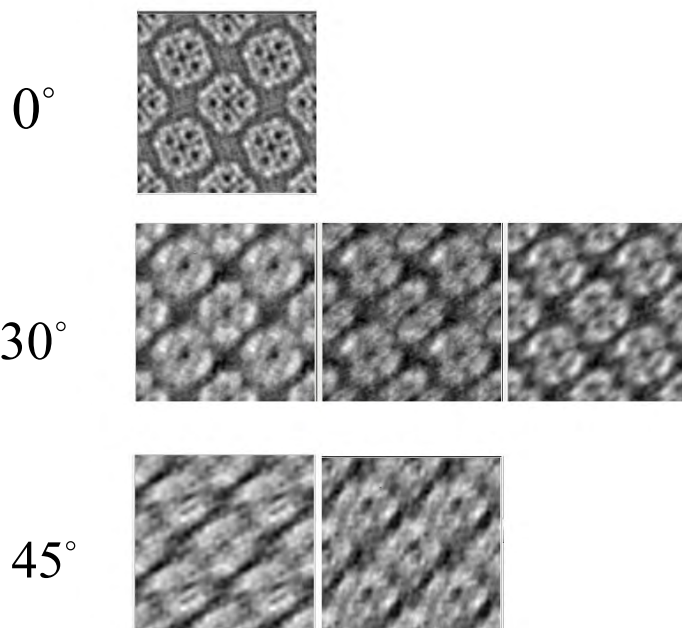
Reference volume  $A$  and particle  $X$ :

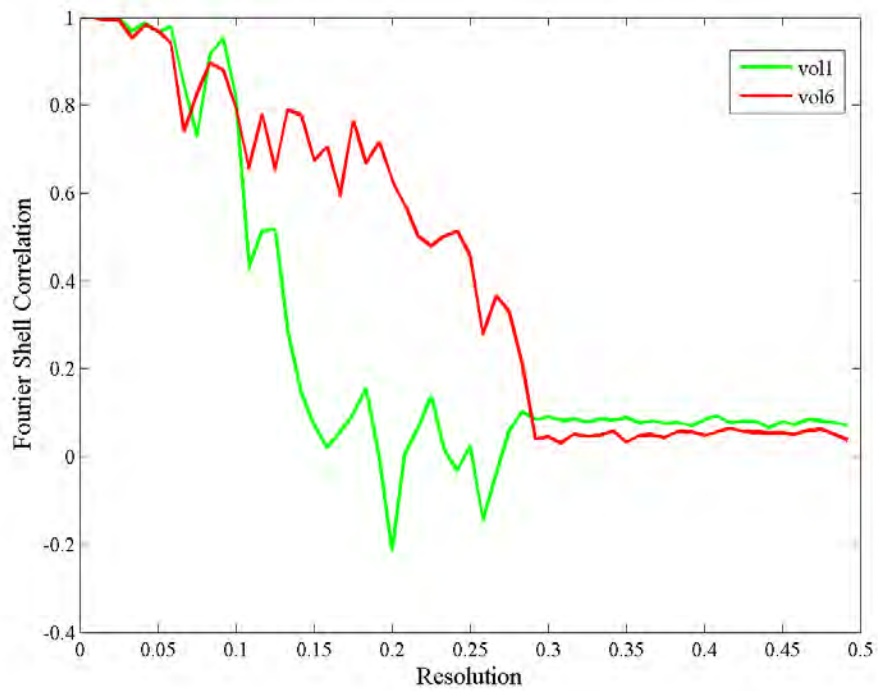
$$\phi_i^{(n+1)} = \arg \max_{\phi} \{X_i \cdot P_{\phi}(A)^{(n)}\} \quad (1)$$

$$A^{(n+1)} = \frac{1}{N} \sum_{i=1}^N I_{\phi}(X_i)^{(n+1)} \quad (2)$$

### 3D Reconstruction Results:

glycerol facilitator of *Escherichia coli*





**Heuristic Optimization Algorithms:** locally search for the translation and orientation parameters

$$\phi = (\varphi, \theta, \psi, x, y)$$

- Powell minimization search in Frealign
- Simulated annealing search

## Using Large Number of Particles

- Accuracy of individual particle is trivial
- SNR is determined by alignment accuracy of large population of particles

The importance of data statistics

## Maximum Likelihood (ML) Approach

- No alignment set  $\phi_i = (\varphi_i, \theta_i, \psi_i, x_i, y_i)$  is determined. It is treated as random variables with some statistical distribution.
- Find the statistical parameters which best describe the observed data.



**Method:** Find the most likely parameters  $\Theta = (A, \sigma, \phi)$  which results in the observed particles

**Noise:** Gaussian distribution with zero mean and std

**Translation and orientation:** Gaussian distribution with zero mean and std

### 3D ML for 2D Crystals

Each particle contributes to all **possible** projections with a probability:

$$\gamma_i(\phi; \Theta) = \left( \frac{1}{\sqrt{2\pi\sigma_k}} \right)^M \exp \left( - \sum_k \frac{|X_i - CTF_i \cdot P_\phi(A)|^2}{2\sigma_k^2} \right) \times pdf(\phi | \Theta)$$

Quality of particles

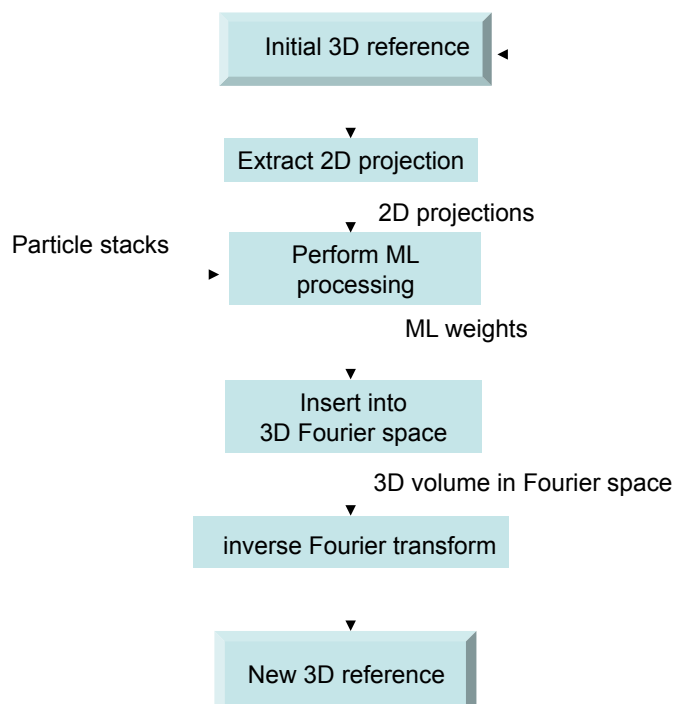
## Advantages of ML

- Free choice of initial reference, robust to low SNR.
- In addition to the weight representing the quality of the images, *pdf* of angles and shifts dominates the probability when SNR is very low or when reference is wrong.

The assumption of angles distribution is critical for the success of ML.

□

## ML 3D Reconstruction



## 3D Reconstruction Results

### **Conclusion and Future Work**

- Single particle approaches can improve the resolution by dealing with the unflatness problem of sample support
- ❖ Explore different 3D alignment approaches
- ❖ Investigate the errors introduced by overlapping projections and develop algorithms to reduce the errors
- ❖ Incorporate the algorithms into 2dx software package

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