# Modifications to the EMC algorithm for orientation recovery in Single Particle Imaging experiments on X-ray free electron lasers

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# X-ray Free Electron Lasers

Key properties:

- Very bright radiation
- Extremely short pulses (10 fs)
- High repetition rate



SwissFEL (2016)





POHANG ACCELERATOR LABORATORY

LCLS (2009)





"Diffraction before destruction" principle takes care of radiation damage problem.

Destroying the sample however, makes it necessary to use many identical samples that enter the beam in an unknown orientation.

Orientation recovery problem

# **EMC** algorithm



Iteration t:

Expansion

Maximization

### Compression

Convert 3d density to tomographic representation  $W_j$  - points corresponding to image in *j-th* orientation. Convert  $W_j$  back to regular grid and impose Friedel symmetry.

$$W_j^t \longrightarrow W_j^{t+1}$$

# Maximization step (EM)

**Expectation:** for each image  $D_i$  compute probability of being in *j-th* orientation  $P_{ij}(W_j^t, D_i)$ 

**Maximization:** compute new densities so that total likelihood is maximized  $A_i^{t+1}$ 

$$W_j^{t+1}(P_j, D) = \frac{\sum_i P_j}{\sum_i P_j}$$

# Online EM

**Expectation:** calculate the probabilities only for 1 image  $P_{i^*j}(W_j^t, D_{i^*})$ 

**Maximization:** compute new densities so that total likelihood is maximized, based on only 1 set of updated probabilities  $W_j^{t+1}(P_{i^*j}, D)$ 

### **Incremental EMC**

Reuse the old probabilities  $P_{ij}^{old}$  for  $i\neq i^{\ast}$ 

$$W_{j}^{t+1} = \frac{A_{j}^{t} + P_{i^{*}j}D_{i^{*}} - P_{i^{*}j}^{old}D_{i^{*}}}{B_{j}^{t} + P_{i^{*}j} - P_{i^{*}j}^{old}}$$

### Stepwise EMC

Interpolate between old  $W_j$  and a new one calculated only from  $P_{i^*j}(W_j^t, D_{i^*})$ 

$$W_j^{t+1} = \frac{(1 - \eta_t)A_j^t + \eta_t P_{i^*j}D_{i^*}}{(1 - \eta_t)B_j^t + \eta_t P_{i^*j}}$$

# Adaptive over-relaxed EMC

### Maximization step:

 $W_j^t \longrightarrow W_j^{t+1}$ Instead of using  $W_j^{t+1}$ , which maximizes total likelihood, we go in that "direction" further:

$$\overline{W_j^{t+1}} = W_j^t + \alpha (W_j^{t+1} - W_j^t), \ \alpha > 1$$

Where coefficient  $\alpha$  is adaptively changed:

If  $\overline{W_j^{t+1}}$  increases likelihood compared to  $W_j^t$ , then we increase  $\alpha$ , otherwise we set  $\alpha$  to 1 and use  $W_j^{t+1}$  instead of  $\overline{W_j^{t+1}}$  as our current approximation.

# Testing the algorithms

### Generating data





1380 images 50 (10x5) iterations

Iterations	1-10	11-20	21-30	31-40	41-50
# of orientations	420	1380	3240	6300	10360

Assessing the results - RMS difference between initial model and the reconstruction.

## RMS difference between model and reconstructions



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# Thanks for your attention!